

CENTRAL LIMIT THEOREMS FOR LINEAR STATISTICS OF HEAVY TAILED RANDOM MATRICES

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ABSTRACT. We show central limit theorems (CLT) for the linear statistics of symmetric matrices with independent heavy tailed entries, including entries in the domain of attraction of α -stable laws and entries with moments exploding with the dimension, as in the adjacency matrices of Erdős-Rényi graphs. For the second model, we also prove a central limit theorem of the moments of its empirical eigenvalues distribution. The limit laws are Gaussian, but unlike the case of standard Wigner matrices, the normalization is the one of the classical CLT for independent random variables.

1. INTRODUCTION AND STATEMENT OF RESULTS

Recall that a Wigner matrix is a symmetric random matrix $A = (a_{i,j})_{i,j=1,\dots,N}$ such that

1. the sub-diagonal entries of A are independent and identically distributed (i.i.d.),
2. the random variables $\sqrt{N}a_{i,j}$ are distributed according to a measure μ that does not depend on N and have all moments finite.

This model was introduced in 1956 by Wigner [48] who proved the convergence of the moments

$$(1) \quad \lim_{N \rightarrow \infty} \mathbb{E} \left[\frac{1}{N} \text{Tr}(A^p) \right] = \int x^p \frac{\sqrt{4-x^2}}{2\pi} dx$$

when μ is centered with unit variance. Moments can be easily replaced by bounded continuous functions in the above convergence and this convergence holds almost surely. Assumption 2 can also be weakened to assume only that the second moment is finite. The fluctuations around this limit or around the expectation were first studied by Jonsson [33]

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in the (slightly different) Wishart model, then by Pastur *et al.* in [34], Sinai and Soshnikov [44] with $p \ll N^{1/2}$ possibly going to infinity with N . Since then, a long list of further-reaching results have been obtained: the central limit theorem was extended to so-called matrix models where the entries interact via a potential in [32], the set of test functions was extended and the assumptions on the entries of the Wigner matrices weakened in [7, 6, 36, 42], a more general model of band matrices was considered in [2] (see also [36, 5] for general covariance matrices), unitary matrices were considered in [31, 23, 45, 22], and Chatterjee developed a general approach to these questions in [20], under the condition that the law μ can be written as a transport of the Gaussian law. Finally, but this is not really our concern here, the fluctuations of the trace of words in several random matrices were studied in [19, 29, 39, 30]. It turns out that in these cases

$$\mathrm{Tr}(A^p) - \mathbb{E}[\mathrm{Tr}(A^p)]$$

converges towards a Gaussian variable whose covariance depends on the first four moments of μ . Moments can also be replaced by regular enough functions and Assumption 2 can be weakened to assume that the fourth moment only is finite. The latter condition is however necessary as the covariance for the limiting Gaussian depends on it. The absence of normalization by \sqrt{N} shows that the eigenvalues of A fluctuate very little, as precisely studied by Erdős, Schlein, Yau, Tao, Vu and their co-authors, who analyzed their rigidity in e.g. [25, 26, 46].

In this article, we extend these results for a variation of the Wigner matrix model where Assumption 2 is removed: some entries of the matrix can be very large, e.g. when μ does not have any second moment or when it depends on N , with moments growing with N . Then, Wigner's convergence theorem (1) does not hold, even when moments are replaced by smooth bounded functions. The analogue of the convergence (1) was studied when the common law μ of the entries of A belongs to the domain of attraction of an α -stable law or μ depends on N and has moments blowing up with N . Although technical, the model introduced in Hypothesis 1.1 below has the advantage of containing these two examples (for u, v some sequences depending implicitly on N , $u \ll v$ means that $u/v \rightarrow 0$ as $N \rightarrow \infty$).

Hypothesis 1.1. *Let, for each $N \geq 1$, $A_N = [a_{ij}]$ be an $N \times N$ real symmetric random matrix whose sub-diagonal entries are some i.i.d. copies of a random variable a (depending implicitly on N) such that:*

- *The random variable a can be decomposed into $a = b + c$ such that as $N \rightarrow \infty$,*

$$(2) \quad \mathbb{P}(c \neq 0) \ll N^{-1}$$

$$(3) \quad \mathrm{Var}(b) \ll N^{-1/2}$$

Moreover, if the b_i 's are independent copies of b ,

$$(4) \quad \lim_{K \rightarrow \infty} \lim_{N \rightarrow \infty} \mathbb{P} \left(\sum_{i=1}^N (b_i - \mathbb{E}(b_i))^2 \geq K \right) = 0.$$

- For any $\varepsilon > 0$ independent of N , the random variable a can be decomposed into $a = b_\varepsilon + c_\varepsilon$ such that

$$(5) \quad \limsup_{N \rightarrow \infty} N \mathbb{P}(c_\varepsilon \neq 0) \leq \varepsilon$$

for all $k \geq 1$, $N\mathbb{E}[(b_\varepsilon - \mathbb{E}b_\varepsilon)^{2k}]$ has a finite limit $C_{\varepsilon,k}$ as $N \rightarrow \infty$.

- For ϕ_N the function defined on the closure $\overline{\mathbb{C}^-}$ of $\mathbb{C}^- := \{\lambda \in \mathbb{C}; \Im \lambda < 0\}$ by

$$(6) \quad \phi_N(\lambda) := \mathbb{E}[\exp(-i\lambda a^2)],$$

we have the convergence, uniform on compact subsets of $\overline{\mathbb{C}^-}$,

$$(7) \quad N(\phi_N(\lambda) - 1) \longrightarrow \Phi(\lambda),$$

for a certain function Φ defined on $\overline{\mathbb{C}^-}$.

Examples of random matrices satisfying Hypothesis 1.1 are defined as follows.

Definition 1.2 (Models of symmetric heavy tailed matrices with i.i.d. sub-diagonal entries).

Let $A = (a_{i,j})_{i,j=1,\dots,N}$ be a random symmetric matrix with i.i.d. sub-diagonal entries.

1. We say that A is a **Lévy matrix** of parameter α in $]0, 2[$ when $A = X/a_N$ where the entries x_{ij} of X have absolute values in the domain of attraction of α -stable distribution, more precisely

$$(8) \quad \mathbb{P}(|x_{ij}| \geq u) = \frac{L(u)}{u^\alpha}$$

with a slowly varying function L , and

$$a_N = \inf\{u : P(|x_{ij}| \geq u) \leq \frac{1}{N}\}$$

($a_N = \tilde{L}(N)N^{1/\alpha}$, with $\tilde{L}(\cdot)$ a slowly varying function).

2. We say that A is a **Wigner matrix with exploding moments** with parameter $(C_k)_{k \geq 1}$ whenever the entries of A are centered, and for any $k \geq 1$

$$(9) \quad N\mathbb{E}[(a_{ij})^{2k}] \xrightarrow{N \rightarrow \infty} C_k,$$

with for a constant $C > 0$. We assume that there exists a unique measure m on \mathbb{R}^+ such that for all $k \geq 0$, $C_{k+1} = \int x^k dm(x)$.

Lemma 1.3. Both Lévy matrices and Wigner matrices with exploding moments satisfy Hypothesis 1.1. For Lévy matrices, the function Φ is given by formula

$$(10) \quad \Phi(\lambda) = -\sigma(i\lambda)^{\alpha/2}$$

for some constant $\sigma \geq 0$ (in this text, as specified in the notations paragraph at the end of this section, the power functions have a cut on \mathbb{R}^-), whereas for Wigner matrices with exploding moments, the function Φ is given by

$$(11) \quad \Phi(\lambda) = \int \underbrace{\frac{e^{-i\lambda x} - 1}{x}}_{:= -i\lambda \text{ for } x = 0} dm(x),$$

for m the unique measure on \mathbb{R}_+ with moments $\int x^k dm(x) = C_{k+1}$, $k \geq 0$.

The proof of this lemma, and of Lemmas 1.8 and 1.12, which show that our hypotheses hold for both Lévy matrices and Wigner matrices, are given in Section 6.

One can easily see that our results also apply to complex Hermitian matrices: in this case, one only needs to require Hypothesis 1.1 to be satisfied by the absolute value of non diagonal entries and to have a_{11} going to zero as $N \rightarrow \infty$.

A Lévy matrix whose entries are truncated in an appropriate way is a Wigner matrix with exploding moments [10, 38, 49]. The recentered version¹ of the adjacency matrix of an Erdős-Rényi graph, *i.e.* of a matrix A such that

$$(12) \quad A_{ij} = 1 \text{ with probability } p/N \text{ and } 0 \text{ with probability } 1 - p/N,$$

is also an exploding moments Wigner matrix, with $\Phi(\lambda) = p(e^{-i\lambda} - 1)$ (the measure m of Lemma 1.3 is $p\delta_1$). In this case the fluctuations were already studied in [43]. The method of [43] can be adapted to study the fluctuations of linear statistics of Wigner matrices with exploding moments. Nevertheless, since we actually use Wigner matrices with exploding moments to study of the fluctuations of Lévy matrices, it is worthwhile to study these ensembles together.

The weak convergence of the empirical eigenvalues distribution of a Lévy matrix has been established in [10] (see also [18, 8, 15]) where it was shown that for any bounded continuous function f ,

$$\lim_{N \rightarrow \infty} \frac{1}{N} \text{Tr}(f(A)) = \int f(x) d\mu_\alpha(x) \quad a.s.$$

where μ_α is a heavy tailed probability measure which depends only on α . Moreover, μ_α converges towards the semicircle law as α goes to 2.

The convergence in moments, in expectation and in probability, of the empirical eigenvalues distribution of a Wigner matrix with exploding moments has been established by Zakharevich in [49]. In that case, moments are well defined and for any continuous bounded function f ,

$$\lim_{N \rightarrow \infty} \frac{1}{N} \text{Tr}(f(A)) = \int f(x) d\mu_{\mathbf{C}}(x) \quad a.s.$$

where $\mu_{\mathbf{C}}$ is a probability measure which depends only on the sequence $\mathbf{C} := (C_k)_{k \geq 1}$.

¹The recentering has in fact asymptotically no effect on the spectral measure A as it is a rank one perturbation.

We shall first state the fluctuations of moments of a Wigner matrix with exploding moments around their limit, namely prove the following theorem.

Theorem 1.4 (The CLT for moments of Wigner matrices with exploding moments).

Let $A = (a_{i,j})_{i,j=1,\dots,N}$ be a Wigner matrix with exploding moments with parameter $(C_k)_{k \geq 1}$. Then the process

$$(13) \quad \left(\frac{1}{\sqrt{N}} \text{Tr } A^K - \mathbb{E} \left[\frac{1}{\sqrt{N}} \text{Tr } A^K \right] \right)_{K \geq 1}$$

converges in distribution to a centered Gaussian process.

This theorem has been established for the slightly more restrictive model of adjacency matrices of weighted Erdős-Rényi graphs in [40, 35, 47]. Our proof is based on the moment method, the covariance of the process is of combinatorial nature, given in Section 2, Formula (40) and Theorem 2.2.

Note that the speed of the central limit theorem is $N^{-1/2}$ as for independent integrable random variables, but differently from what happens for standard Wigner's matrices. This phenomenon has also already been observed for adjacency matrix of random graphs [9, 24] and we will see below that it also holds for Lévy matrices. It suggests that the repulsive interactions exhibited by the eigenvalues of most models of random matrices with lighter tails than heavy tailed matrices no longer work here.

For Lévy matrices, moments do not make sense and one should consider smooth bounded test functions. We start as is common in random matrix theory, with the study of the normalized trace of the resolvent of A , given for $z \in \mathbb{C} \setminus \mathbb{R}$ by

$$G(z) := \frac{1}{z - A}.$$

By the previous results, for both Lévy and exploding moments Wigner matrices, $N^{-1} \text{Tr } G(z)$ converges in probability to a deterministic limit as the parameter N tends to infinity. We study the associated fluctuations.

In fact, even in the case of Wigner matrices with exploding moments, the CLT for moments does not imply a priori the CLT for Stieltjes functions even though concentration inequalities hold on the right scale, see [16]. Indeed, one cannot approximate smooth bounded functions by polynomials for the total variation norm unless one can restrict oneself to compact subsets, a point which is not clear in this heavy tail setting. However, with additional arguments based on martingale technology, we shall prove the following result, valid for both Lévy matrices and Wigner matrices with exploding moments.

Theorem 1.5 (CLT for Stieltjes transforms). *Under Hypothesis 1.1, where we assume additionally that for $\varepsilon > 0$, there exists $C_\varepsilon > 0$ such that, for any $k \geq 1$, one has*

$$(14) \quad \lim_{N \rightarrow \infty} N \mathbb{E}[(b_\varepsilon - \mathbb{E} b_\varepsilon)^2] = C_{\varepsilon,k} \leq C_\varepsilon^k$$

then, the process

$$(15) \quad \left(\frac{1}{\sqrt{N}} \operatorname{Tr} G(z) - \frac{1}{\sqrt{N}} \mathbb{E}[\operatorname{Tr} G(z)] \right)_{z \in \mathbb{C} \setminus \mathbb{R}}$$

converges in distribution (in the sense of finite marginals) to a centered complex Gaussian process $Z(z)$, $z \in \mathbb{C} \setminus \mathbb{R}$, whose covariance $\operatorname{Cov}(Z(z), Z(z')) := \mathbb{E}[Z(z)Z(z')] = C(z, z')$ is given in Formulas (17) (18) and (19) below.

Note that the uniform convergence on compact subsets, in (7), implies that Φ is analytic on \mathbb{C}^- and continuous on $\overline{\mathbb{C}^-}$. As a consequence, we can extend the central limit theorem to a larger class of functions, namely the Sobolev space $W_0^{\infty,1}(\mathbb{R})$ of absolutely continuous functions with null limits at $\pm\infty$ and with finite total variation. It is easy to see that $W_0^{\infty,1}(\mathbb{R})$ is the set of functions $f : \mathbb{R} \rightarrow \mathbb{R}$ such that there is $g \in L^1(\mathbb{R})$ such that

$$\int g(t)dt = 0 \quad \text{and} \quad \forall x \in \mathbb{R}, f(x) = \int_{-\infty}^x g(t)dt.$$

We endow $W_0^{\infty,1}(\mathbb{R})$ with the total variation norm defined at (100) (see also (101)) in Section 7.1 of the appendix. It is easy to see that when restricted to $W_0^{\infty,1}(\mathbb{R})$, this norm is equivalent to the one usually used on the (larger) space $W^{\infty,1}(\mathbb{R})$ (see [17, Sec. 8.2]) and that the set $\mathcal{C}_c^2(\mathbb{R})$ of \mathcal{C}^2 real valued functions with compact support on \mathbb{R} is dense in $W_0^{\infty,1}(\mathbb{R})$.

Corollary 1.6. *Under the hypotheses of Theorem 1.5, the process*

$$(16) \quad \left(Z_N(f) := \frac{1}{\sqrt{N}} \operatorname{Tr} f(A) - \frac{1}{\sqrt{N}} \mathbb{E}[\operatorname{Tr} f(A)] ; f \in W_0^{\infty,1}(\mathbb{R}) \right)$$

converges in law towards a centered Gaussian process $(Z(f) ; f \in W_0^{\infty,1}(\mathbb{R}))$ with covariance given by the unique continuous extension to $W_0^{\infty,1}(\mathbb{R}) \times W_0^{\infty,1}(\mathbb{R})$ of the functional defined on $\mathcal{C}_c^2(\mathbb{R}) \times \mathcal{C}_c^2(\mathbb{R})$ by

$$C(f, g) = \mathbb{E} \left[\Re \left(\int_{-\infty}^{\infty} dx \int_0^{+\infty} dy \bar{\partial} \Psi_f(x, y) Z(x + iy) \right) \Re \left(\int_{-\infty}^{\infty} dx \int_0^{+\infty} dy \bar{\partial} \Psi_g(x, y) Z(x + iy) \right) \right]$$

where Z is a centered gaussian process with covariance $C(z, z')$ given in Theorem 1.5 and the function Ψ_f is given by

$$\Psi_f(x, y) = f(x)\phi(y) + if'(x)\phi(y)y, \quad \bar{\partial} \Psi(x, y) = \pi^{-1}(\partial_x + i\partial_y)\Psi(x, y)$$

where ϕ is a smooth compactly supported function equal to one in the neighborhood of the of the origin.

The function $C(z, z')$ in Theorem 1.5 is given by $C(z, z') = L(z, z') - L(z)L(z')$, where

$$(17) \quad L(z) = \int_0^{\operatorname{sgn} \Im(z)\infty} \frac{1}{t} \partial_z e^{itz + \rho_z(t)} dt,$$

$$(18) \quad L(z, z') = \int_0^1 \int_0^{\operatorname{sgn} \Im(z)\infty} \int_0^{\operatorname{sgn} \Im(z')\infty} \frac{1}{tt'} \partial_{z,z'}^2 e^{itz + it'z' + \rho_{z,z'}^u(t,t')} dt dt' du.$$

The maps ρ_z and $\rho_{z,z'}^u$ are given by $\rho_{z,z'}^u(t, t') = \mathbb{E}_{z,z'}^u[\Phi(ta + t'a')]$, $\rho_z(t) = \rho_{z,z'}^u(t, 0)$, where (a, a') is distributed according to the non random weak limit as $N \rightarrow \infty$, $k/N \rightarrow u$ of

$$(19) \quad \frac{1}{N-1} \sum_{j=1}^{N-1} \delta_{G_k(z)_{jj}, G'_k(z')_{jj}},$$

where $G_k(z) = (z - A_k)^{-1}$ for A_k the matrix obtained from A by deleting the k -th row and column, and $G'_k(z') = (z' - A'_k)^{-1}$ for A'_k a copy of A_k where the entries (i, j) for i or $j \geq k$ are independent of A_k and the others are those of A_k . Also, we assume $t\Im z \geq 0$ and $t'\Im z' \geq 0$ in (19).

The existence of the limit (19) is a consequence of a generalized convergence in moments, namely the convergence in distribution of traffics, of (A_k, A'_k) stated in [38], see Lemma 3.2. However, under stronger assumptions, an independent proof of this convergence and an intrinsic characterization of $\rho_{z,z'}^u$ are provided in Theorem 1.13 below.

Let us first mention that the map ρ_z is the almost sure point-wise limit

$$(20) \quad \rho_z(t) = \lim_{N \rightarrow \infty} \frac{1}{N-1} \sum_{j=1}^{N-1} \Phi(tG_k(z)_{jj}),$$

and it characterizes the limiting eigenvalues distribution of A . Indeed, under the assumptions of Theorem 1.5, for all $z \in \mathbb{C}^+$, we have the almost sure convergence

$$(21) \quad \lim_{N \rightarrow \infty} \frac{1}{N} \text{Tr } G_N(z) = i \int_0^\infty e^{itz + \rho_z(t)} dt.$$

This fact was known in the Lévy case [10] and is proved in greater generality in Corollary 3.3. Let us first give a characterization of ρ_z .

Hypothesis 1.7. *The function Φ of (7) admits the decomposition*

$$(22) \quad \Phi(z) = \int_0^\infty g(y) e^{i\frac{y}{z}} dy$$

where $g(y)$ is a function such that for some constants $K, \gamma > -1, \kappa \geq 0$, we have

$$(23) \quad |g(y)| \leq K \mathbb{1}_{y \leq 1} y^\gamma + K \mathbb{1}_{y \geq 1} y^\kappa, \quad \forall y > 0.$$

The following lemma insures us that our two main examples satisfy Hypothesis 1.7 (note that it is also the case when the function $\Phi(x^{-1})$ is in L^1 and has its Fourier transform supported by \mathbb{R}_+).

Lemma 1.8. *Both Lévy matrices and Wigner matrices with exploding moments satisfy Hypothesis 1.7. For Lévy matrices, the function g is $g(y) = C_\alpha y^{\frac{\alpha}{2}-1}$, with $C_\alpha = -\sigma i^{\alpha/2}$,*

whereas for Wigner matrices with exploding moments, the function g is

$$(24) \quad g(y) = - \int_{\mathbb{R}^+} \underbrace{\frac{J_1(2\sqrt{xy})}{\sqrt{xy}}}_{:=1 \text{ for } xy=0} dm(x),$$

for m the measure on \mathbb{R}_+ of Lemma 1.3 and J_1 the Bessel function of the first kind defined by $J_1(s) = \frac{s}{2} \sum_{k \geq 0} \frac{(-s^2/4)^k}{k!(k+1)!}$.

Theorem 1.9 (A fixed point equation for $\rho_z(t)$). *Under Hypotheses 1.1 and 1.7, the function $(z \in \mathbb{C}^+, \lambda \in \mathbb{R}^+) \mapsto \rho_z(\lambda)$ is analytic in its first argument and continuous in the second, with non positive real part, and characterized among the set of such functions by the formula*

$$(25) \quad \rho_z(\lambda) = \lambda \int_0^\infty g(\lambda y) e^{iyz + \rho_z(y)} dy.$$

Above, the second point in Hypotheses 1.1 is not required anymore, as it served mainly to prove convergence, which is now insured by the uniqueness of limit points.

Note that the asymptotics of Wigner matrices with bounded moments is also described by (25). In this case $\Phi(\lambda) = -i\lambda$, so $g(y) = -i$ and $\rho_z(t) = -it \lim_{N \rightarrow \infty} \frac{1}{N} \text{Tr } G(z)$, which leads, by formula (21), to the classical quadratic equation

$$(26) \quad s(z) := \lim_{N \rightarrow \infty} \frac{1}{N} \text{Tr } G(z) = \frac{1}{z - s(z)}.$$

Let us now give a fixed point characterization for the function $\rho_{z,z'}(t, t')$ of (19).

Hypothesis 1.10. *The function Φ of (7) either has the form*

$$(27) \quad \Phi(x) = -\sigma(ix)^{\alpha/2}$$

or admits the decomposition, for x, y non zero:

$$(28) \quad \Phi(x+y) = \iint_{(\mathbb{R}_+)^2} e^{i\frac{v}{x} + i\frac{v'}{y}} d\tau(v, v') + \int e^{i\frac{v}{x}} d\mu(v) + \int e^{i\frac{v'}{y}} d\mu(v')$$

for some complex measures τ, μ on respectively $(\mathbb{R}_+)^2$ and \mathbb{R}_+ such that for all $b > 0$, $\int e^{-bv} d|\mu|(v)$ is finite and for some constants $K > 0$, $-1 < \gamma \leq 0$ and $\kappa \geq 0$, and

$$(29) \quad \frac{d|\tau|(v, v')}{dv dv'} \leq K (v^\gamma \mathbb{1}_{v \in [0,1]} + v^\kappa \mathbb{1}_{v \in [1,\infty[}) (v'^\gamma \mathbb{1}_{v' \in [0,1]} + v'^\kappa \mathbb{1}_{v' \in [1,\infty[}).$$

Remark 1.11. For Lévy matrices, where $\Phi(x) = C_\alpha x^{\alpha/2}$ with $C_\alpha = -\sigma i^{\alpha/2}$, (28) holds as well. Indeed, for all $x, y \in \mathbb{C}^+$ (with a constant C_α that can change at every line),

$$(30) \quad \begin{aligned} \Phi(x^{-1} + y^{-1}) &= C_\alpha \left(\frac{1}{x} + \frac{1}{y} \right)^{\alpha/2} = C_\alpha \frac{1}{x^{\alpha/2}} \frac{1}{y^{\alpha/2}} (x + y)^{\alpha/2} \\ &= C_\alpha \int_0^\infty dw \int_0^\infty dw' \int_0^\infty dv w^{\alpha/2-1} (w')^{\alpha/2-1} v^{-\alpha/2-1} e^{iwx+iw'y} (e^{iv(x+y)} - 1) \end{aligned}$$

(where we used the formula $z^{\alpha/2} = C_\alpha \int_{t=0}^{+\infty} \frac{e^{itz} - 1}{t^{\alpha/2+1}} dt$ for any $z \in \mathbb{C}^+$ and $\alpha \in (0, 2)$, which can be proved with the residues formula) so that (28) holds with $\mu = 0$ and $\tau(v, v')$ with density with respect to Lebesgue measure given by

$$(31) \quad C_\alpha \int_0^{+\infty} u^{-\alpha/2-1} \{ (v-u)^{\alpha/2-1} (v'-u)^{\alpha/2-1} \mathbb{1}_{0 \leq u \leq v \wedge v'} - v^{\alpha/2-1} v'^{\alpha/2-1} \} du.$$

Unfortunately τ does not satisfy (29) as its density blows up at $v = v'$: we shall treat both case separately.

The following lemma insures us that our two main examples satisfy Hypothesis 1.10.

Lemma 1.12. *Both Lévy matrices and Wigner matrices with exploding moments satisfy Hypothesis 1.10. For Wigner matrices with exploding moments, the measure τ is given by*

$$(32) \quad d\tau(v, v') := dv dv' \int \frac{J_1(2\sqrt{vx}) J_1(2\sqrt{v'x})}{\sqrt{vv'}} dm(x)$$

for m the measure on \mathbb{R}_+ of Lemma 1.3 and the measure μ is given by

$$(33) \quad d\mu(v) := -dv \int \frac{J_1(2\sqrt{vx})}{\sqrt{v}} dm(x).$$

Theorem 1.13 (A fixed point system of equations for $\rho_{z,z'}^u(t, t')$). *Under Hypotheses 1.1, 1.7 and 1.10, the conclusions of Theorem 1.5 and Corollary 1.6 hold and the parameter $\rho_{z,z'}$ of (18) is given by*

$$\rho_{z,z'}^u(t, s) = u \rho_{z,z'}^{u,1}(t, s) + (1-u) \rho_{z,z'}^2(t, s)$$

where $\rho_{z,z'}^{u,1}(t, s), \rho_{z,z'}^2(t, s)$ are analytic functions on $\Lambda = \{z, z' : t\Im z > 0, s\Im z' > 0\}$ and uniformly continuous on compacts in the variables (t, s) (and β -Hölder for $\beta > \alpha/2$ in the Lévy matrices case, see Lemma 5.1), given by (35) as far as $\rho_{z,z'}^{u,2}(t, s)$ is concerned and unique solution, among such functions, of the following fixed point equation (34) as far as

$\rho_{z,z'}^{u,1}(t, s)$ is concerned:

$$(34) \quad \rho_{z,z'}^{u,1}(t, s) = \int_0^{\operatorname{sgn}_z \infty} \int_0^{\operatorname{sgn}_{z'} \infty} e^{i\frac{v}{t}z + i\frac{v'}{s}z'} e^{u\rho_{z,z'}^{u,1}(\frac{v}{t}, \frac{v'}{s}) + (1-u)(\rho_z(\frac{v}{t}) + \rho_{z'}(\frac{v'}{s}))} d\tau(v, v') \\ + \int_0^{\operatorname{sgn}_z \infty} e^{i\frac{v}{t}} e^{\rho_z(\frac{v}{t})} d\mu(v) + \int_0^{\operatorname{sgn}_{z'} \infty} e^{i\frac{v'}{s}} e^{\rho_{z'}(\frac{v'}{s})} d\mu(v),$$

$$(35) \quad \rho_{z,z'}^2(t, s) = \int_0^{\operatorname{sgn}_z \infty} \int_0^{\operatorname{sgn}_{z'} \infty} e^{i\frac{v}{t}z + i\frac{v'}{s}z'} e^{\rho_z(\frac{v}{t}) + \rho_{z'}(\frac{v'}{s})} d\tau(v, v') \\ + \int_0^{\operatorname{sgn}_z \infty} e^{i\frac{v}{t}} e^{\rho_z(\frac{v}{t})} d\mu(v) + \int_0^{\operatorname{sgn}_{z'} \infty} e^{i\frac{v'}{s}} e^{\rho_{z'}(\frac{v'}{s})} d\mu(v),$$

with the notations $\operatorname{sgn}_z := \operatorname{sign}(\Im z)$, $\operatorname{sgn}_{z'} := \operatorname{sign}(\Im z')$ and the measures τ, μ defined by Hypothesis 1.10 and Remark 1.11.

Let us conclude this introduction with three remarks.

- (1) Let $A = X/a_N$ be a Lévy matrix as defined at Definition 1.2 but with $\alpha = 2$. Then using Example c) p. 44. of [27] instead of the hypothesis made at Equation (7), one can prove that as $N \rightarrow \infty$, the spectral measure of A converges almost surely to the semi-circle law with support $[-2, 2]$ (see (26)). This result somehow “fills the gap” between heavy-tailed matrices and finite second moment Wigner matrices. It allows for example to state that if $P(|X_{ij}| \geq u) \sim cu^{-2}$, with $c > 0$, even though the entries of X do not have any second moment, we have that the empirical spectral law of $\frac{X}{\sqrt{cN \log(N)}}$ converges almost surely to the semi-circle law with support $[-2, 2]$.
- (2) Our results also have an application to **standard Wigner matrices** (*i.e.* symmetric random matrices of the form $A = X/\sqrt{N}$, with X having centered i.i.d. sub-diagonal entries with variance one and not depending on N). In this case, the function Φ of (7) is linear, which implies that $L(z, z') = L(z)L(z')$ for all z, z' , so that the covariance is null, (25) is the self-consistent equation satisfied by the Stieltjes transform of the semi-circle law, namely (26), and Corollary 1.6 only means that for functions $f \in \mathcal{A}$, we have, for the convergence in probability,

$$(36) \quad \operatorname{Tr} f(A) - \mathbb{E}[\operatorname{Tr} f(A)] = o(\sqrt{N}).$$

This result is new for Wigner matrices whose entries have a second but not a fourth moment, (36) brings new information. Indeed, for such matrices, which could be called “*semi heavy-tailed random matrices*”, the convergence to the semi circle law holds (see [4] or the remark right above that one) but the largest eigenvalues do not tend to the upper-bound of the support of the semi-circle law, are asymptotically in the scale $N^{\frac{4-\alpha}{2\alpha}}$ (with $\alpha \in (2, 4)$ as in Equation (8) when such an exponent exists) and distributed according to a Poisson process (see [3]), and it is not clear what the rate of convergence to the semi-circle law will be. Equation (36) shows that this rate is $\ll N^{-1/2}$.

- (3) About recentering with respect to the limit instead of the expectation, it depends on the rate of convergence in (9) or in (6). For instance, if $N\mathbb{E}(a_{11}^{2k}) - C_k = o(N^{-1/2})$ for any $k \geq 1$, then

$$\sqrt{N} \left(\mathbb{E} \left[\frac{1}{N} \text{Tr } A^k \right] - \lim_{N \rightarrow \infty} \mathbb{E} \left[\frac{1}{N} \text{Tr } A^k \right] \right) \xrightarrow[N \rightarrow \infty]{} 0,$$

but otherwise a non trivial recentering should occur. See the end of Section 2.

Organization of the article: The CLTs for moments and Stieltjes transform (Theorems 1.4 and 1.5) are proved in Sections 2 and 3 respectively. Corollary 1.6, which extends the CLTs for functions in $W_0^{\infty,1}$, is proved in Section 4. Theorems 1.9 and 1.13 about fixed point equations for the functions ρ_z and $\rho_{z,z'}^u$ expressing the limit spectral distribution and covariance are proved in Section 5. Section 6 is devoted to the proof of our assumptions for Lévy matrices and Wigner matrices with exploding moments.

Notation: In this article, the power functions are defined on $\mathbb{C} \setminus \mathbb{R}_-$ via the standard determination of the argument on this set taking values in $(-\pi, \pi)$. The set \mathbb{C}^+ (resp. \mathbb{C}^-) denotes the open upper (resp. lower) half plane and for any $z \in \mathbb{C}$, $\text{sgn}_z := \text{sign}(\Im z)$.

2. CLT FOR THE MOMENTS OF WIGNER MATRICES WITH EXPLODING MOMENTS

The goal of this section is to prove Theorem 1.4. In order to prove the CLT for the moments of the empirical eigenvalues distribution of A , we use a modification of the method of moments inspired by [37] which consists of studying more general functionals of the entries of the matrix (the so-called injective moments) than only its moments. We describe this approach below.

Let A be a Wigner matrix with exploding moments. Let $K \geq 1$ be an integer. The normalized trace of the K -th power of A can be expanded in the following way.

$$\begin{aligned} \frac{1}{N} \text{Tr } A^K &= \frac{1}{N} \sum_{i_1, \dots, i_K=1}^N A(i_1, i_2) \dots A(i_{K-1}, i_K) A(i_K, i_1) \\ &= \sum_{\pi \in \mathcal{P}(K)} \underbrace{\frac{1}{N} \sum_{\mathbf{i} \in S_\pi} A(i_1, i_2) \dots A(i_{K-1}, i_K) A(i_K, i_1)}_{\tau_N^0[\pi]}, \end{aligned}$$

where $\mathcal{P}(K)$ is the set of partitions of $\{1, \dots, K\}$ and S_π is the set of multi-indices $\mathbf{i} = (i_1, \dots, i_K)$ in $\{1, \dots, N\}^K$ such that $n \sim_\pi m \Leftrightarrow i_n = i_m$.

We interpret τ_N^0 as a functional on graphs instead of partitions. Let π be a partition of $\{1, \dots, K\}$. Let $T^\pi = (V, E)$ be the undirected graph (with possibly multiple edges and loops) whose set of vertices V is π and with multi-set of edges E given by: there is one

edge between two blocks V_i and V_j of π for each n in $\{1, \dots, K\}$ such that $n \in V_i$ and $n+1 \in V_j$ (with notation modulo K). Then, one has

$$(37) \quad \tau_N^0(\pi) = \tau_N^0[T^\pi]$$

if for graph $T = (V, E)$, we have denoted

$$(38) \quad \tau_N^0(T) = \frac{1}{N} \sum_{\substack{\phi: V \rightarrow [N] \\ \text{injective}}} \prod_{e \in E} A(\phi(e)),$$

where $[N] = \{1, \dots, N\}$ and for any edge $e = \{i, j\}$ we have denoted $A(\phi(e)) = A(\phi(i), \phi(j))$. There is no ambiguity in the previous definition since the matrix A is symmetric.

In order to prove the convergence of

$$(Z_N(K))_{K \geq 1} = \left(\frac{1}{\sqrt{N}} \text{Tr } A^K - \mathbb{E} \left[\frac{1}{\sqrt{N}} \text{Tr } A^K \right] \right)_{K \geq 1}$$

to a Gaussian process, it is sufficient to prove the convergence of

$$(39) \quad (Z_N(T^\pi))_{\pi \in \cup_K \mathcal{P}(K)} := \left(\sqrt{N} \left(\tau_N^0[T^\pi] - \mathbb{E}[\tau_N^0[T^\pi]] \right) \right)_{\pi \in \cup_K \mathcal{P}(K)}$$

to a Gaussian process, since

$$(40) \quad (Z_N(K))_{K \geq 1} = \left(\sum_{\pi \in \mathcal{P}(K)} Z_N(T^\pi) \right)_{K \geq 1}.$$

Before giving the proof of this fact, we recall a result from [38], namely the convergence of $\tau_N^0[T^\pi]$ for any partition π . These limits are involved in our computation of the covariance of the limiting process of $(Z_N(T^\pi))_{\pi \in \cup_K \mathcal{P}(K)}$, and this convergence will be useful in the proof of the CLT for Stieltjes transforms latter.

Proposition 2.1 (Convergence of generalized moments). *Let A be a Wigner matrix with exploding moments with parameter $(C_k)_{k \geq 1}$. For any partition π in $\cup_K \mathcal{P}(K)$, with $\tau_N^0[T^\pi]$ defined in (37),*

$$\mathbb{E}[\tau_N^0[T^\pi]] \xrightarrow{N \rightarrow \infty} \tau^0[T^\pi] := \begin{cases} \prod_{k \geq 1} C_k^{q_k} & \text{if } T^\pi \text{ is a fat tree,} \\ 0 & \text{otherwise,} \end{cases}$$

where a fat tree is a graph that becomes a tree when the multiplicity of the edges is forgotten, and for T such a graph we have denoted q_k the number of edges of T with multiplicity $2k$.

Theorem 2.2 (Fluctuations of generalized moments). *Let A be a Wigner matrix with exploding moments. Then, the process $(Z_N(T^\pi))_{\pi \in \cup_K \mathcal{P}(K)}$ defined by (39) converges to a centered Gaussian process $(z(T^\pi))_{\pi \in \cup_K \mathcal{P}(K)}$ whose covariance is given by: for any T^{π_1}, T^{π_2} ,*

$$\mathbb{E}[z(T^{\pi_1})z(T^{\pi_2})] = \sum_{T \in \mathcal{P}_\#(T^{\pi_1}, T^{\pi_2})} \tau^0[T],$$

where $\tau^0[T]$ is given by Proposition 2.1 and $\mathcal{P}_\#(T^{\pi_1}, T^{\pi_2})$ is the set of graphs obtained by considering disjoint copies of the graphs T^{π_1} and T^{π_2} and gluing them by requiring that they have at least one edge (and therefore two “adjacent” vertices) in common.

Proof. We show the convergence of joint moments of $(Z_N(T^\pi))$. Gaussian distribution being characterized by its moments, this will prove the theorem. Let $T_1 = (V_1, E_1), \dots, T_p = (V_n, E_n)$ be finite undirected graphs, each of them being of the form T^π for a partition π . We first write

$$\begin{aligned} \mathbb{E}[Z_N(T_1) \dots Z_N(T_n)] &= \frac{1}{N^{\frac{n}{2}}} \sum_{\substack{\phi_1, \dots, \phi_n \\ \phi_j: V_j \rightarrow [N] \text{ inj.}}} \underbrace{\mathbb{E} \left[\prod_{j=1}^n \left(\prod_{e \in E_j} A(\phi_j(e)) - \mathbb{E} \left[\prod_{e \in E_j} A(\phi_j(e)) \right] \right) \right]}_{\omega_N(\phi_1, \dots, \phi_n)} \\ &= \sum_{\sigma \in \mathcal{P}(V_1, \dots, V_n)} \frac{1}{N^{\frac{n}{2}}} \sum_{(\phi_1, \dots, \phi_n) \in S_\sigma} \omega_N(\phi_1, \dots, \phi_n), \end{aligned}$$

where

- $\mathcal{P}(V_1, \dots, V_n)$ is the set of partitions of the disjoint union of V_1, \dots, V_n whose blocks contain at most one element of each V_j ,
- S_σ is the set of families of injective maps, $\phi_j : V_j \rightarrow [N], j = 1, \dots, n$, such that for any $v \in V_j, v' \in V_{j'}$, one has $\phi_j(v) = \phi_{j'}(v') \Leftrightarrow v \sim_\sigma v'$.

First, it should be noticed that by invariance in law of A by conjugacy by permutation matrices, for any σ in $\mathcal{P}(V_1, \dots, V_n)$ and (ϕ_1, \dots, ϕ_n) in S_σ , the quantity $\omega_N(\phi_1, \dots, \phi_n)$ depends only on σ . We then denote $\omega_N(\sigma) = \omega_N(\phi_1, \dots, \phi_n)$. Moreover, choosing a partition in $\mathcal{P}(V_1, \dots, V_n)$ is equivalent to merge certain vertices of different graphs among T_1, \dots, T_n . We equip $\mathcal{P}(V_1, \dots, V_n)$ with the edges of T_1, \dots, T_n and say that two vertices are adjacent if there is an edge between them. We denote by $\mathcal{P}_\#(V_1, \dots, V_n)$ the subset of $\mathcal{P}(V_1, \dots, V_n)$ such that any graph has two adjacent vertices that are merged to two adjacent vertices of an other graph. By the independence of the entries of X and the centering of the components in ω_N , for any σ in $\mathcal{P}(V_1, \dots, V_n) \setminus \mathcal{P}_\#(V_1, \dots, V_n)$ one has $\omega_N(\sigma) = 0$. Hence, since the cardinal of S_σ is $\frac{N!}{(N-|\sigma|)!}$, we get

$$\begin{aligned} \mathbb{E}[Z_N(T_1) \dots Z_N(T_n)] &= \sum_{\sigma \in \mathcal{P}_\#(V_1, \dots, V_n)} N^{-\frac{n}{2}} \frac{N!}{(N-|\sigma|)!} \omega_N(\sigma) \\ (41) \quad &= \sum_{\sigma \in \mathcal{P}_\#(V_1, \dots, V_n)} N^{-\frac{n}{2} + |\sigma|} \omega_N(\sigma) (1 + O(N^{-1})). \end{aligned}$$

Let σ in $\mathcal{P}_\#(V_1, \dots, V_n)$. We now analyze the term $\omega_N(\sigma)$. We first expand its product.

$$\omega_N(\sigma) = \sum_{B \subset \{1, \dots, n\}} (-1)^{n-|B|} \mathbb{E} \left[\prod_{j \in B} \prod_{e \in E_j} A(\phi_j(e)) \right] \times \prod_{j \notin B} \mathbb{E} \left[\prod_{e \in E_j} A(\phi_j(e)) \right].$$

Let $B \subset \{1, \dots, n\}$. Denote by T_B the graph obtained by merging the vertices of T_j , $j \in B$ that belong to a same block of σ . For any $k \geq 1$ denote by p_k the number of vertices of T_B where k loops are attached. For any $k \geq \ell \geq 0$, denote by $q_{k,\ell}$ the number of pair of vertices that are linked by k edges in one direction and ℓ edges in the other. Denote by μ_N the common law of the entries of $\sqrt{N}A$. By independence of the entries of A , for any (ϕ_1, \dots, ϕ_n) in S_σ , one has

$$\begin{aligned} \mathbb{E} \left[\prod_{j \in B} \prod_{e \in E_j} A(\phi_j(e)) \right] &= \prod_{k \geq 1} \left(\frac{\int t^k d\mu_N(t)}{N^{\frac{k}{2}}} \right)^{p_k} \prod_{k, \ell \geq 0} \left(\frac{\int t^{k+\ell} d\mu_N(t)}{N^{\frac{k+\ell}{2}}} \right)^{q_{k,\ell}} \\ (42) \quad &= N^{-|\bar{E}_B|} \underbrace{\prod_{k \geq 1} \left(\frac{\int t^k d\mu_N(t)}{N^{\frac{k}{2}-1}} \right)^{p_k} \prod_{k, \ell \geq 0} \left(\frac{\int t^{k+\ell} d\mu_N(t)}{N^{\frac{k+\ell}{2}-1}} \right)^{q_{k,\ell}}}_{\delta_N(B)}, \end{aligned}$$

where $|\bar{E}_B|$ is the number of edges of T_B once the multiplicity and the orientation of edges are forgotten. Recall assumption (9): for any $k \geq 1$,

$$N\mathbb{E}[(a_{i,j})^{2k}] = \mathbb{E} \left[\frac{(\sqrt{N}a_{i,j})^{2k}}{N^{k-1}} \right] = \frac{\int t^{2k} d\mu_N(t)}{N^{k-1}} \xrightarrow{N \rightarrow \infty} C_k.$$

By the Cauchy-Schwarz inequality, for any $k \geq 1$,

$$\frac{\int |t|^{k+1} d\mu_N(t)}{N^{\frac{k+1}{2}-1}} \leq \sqrt{\frac{\int t^{2k} d\mu_N(t)}{N^{k-1}}} \times \sqrt{\int t^2 d\mu_N(t)} = O(1).$$

Hence, since the measure μ_N is centered, the quantity $\delta_N(B)$ is bounded. Denote T_σ the graph obtained by merging the vertices of T_1, \dots, T_n that belong to a same block of σ , $|\bar{E}_\sigma|$ its number of edges when orientation and multiplicity is forgotten, and by c_σ its number of components. We obtain from (41) and (42)

$$\begin{aligned} &\mathbb{E} [Z_N(T_1) \dots Z_N(T_n)] \\ &= \sum_{\sigma \in \mathcal{P}_\#(V_1, \dots, V_n)} \sum_{B \subset \{1, \dots, n\}} N^{-\frac{n}{2} + |\sigma| - |\bar{E}_B| - \sum_{j \notin B} |\bar{E}_{\{j\}}|} (-1)^{n-|B|} \delta_N(B) \prod_{j \notin B} \delta_N(\{j\}) (1 + O(N^{-1})) \\ &= \sum_{\sigma \in \mathcal{P}_\#(V_1, \dots, V_n)} \sum_{B \subset \{1, \dots, n\}} N^{c_\sigma - \frac{n}{2}} \times N^{|\bar{E}_\sigma| - |\bar{E}_B| - \sum_{j \notin B} |\bar{E}_{\{j\}}|} \times N^{|\sigma| - c_\sigma - |\bar{E}_\sigma|} \\ &\quad \times (-1)^{n-|B|} \delta_N(B) \prod_{j \notin B} \delta_N(\{j\}) \times (1 + O(N^{-1})). \end{aligned}$$

A partition $\sigma \in \mathcal{P}_\#(V_1, \dots, V_n)$ induces a partition $\bar{\sigma}$ of $\{1, \dots, n\}$: $i \sim_{\bar{\sigma}} j$ if and only if T_i and T_j belong to a same connected component of T_σ . Denote by $\mathcal{P}_2(n)$ the set of pair partitions of $\{1, \dots, n\}$. One has

$$(43) \quad N^{c_\sigma - \frac{n}{2}} = \mathbb{1}_{\bar{\sigma} \in \mathcal{P}_2(n)} + O(N^{-1}).$$

Secondly, one has $|\bar{E}_\sigma| - |\bar{E}_B| - \sum_{j \notin B} |\bar{E}_{\{j\}}| \leq 0$ with equality if and only if $B = \{1, \dots, n\}$, so that

$$(44) \quad N^{|\bar{E}_\sigma| - |\bar{E}_B| - \sum_{j \notin B} |\bar{E}_{\{j\}}|} = \mathbb{1}_{B=\{1, \dots, n\}} + O(N^{-1}).$$

Moreover, by [28, Lemma 1.1] $|\sigma| - c_\sigma - |\bar{E}_\sigma|$ is the number of cycles of \bar{T}_σ , the graph obtained from T_σ by forgetting the multiplicity and the orientation of its edges. Hence,

$$(45) \quad N^{|\sigma| - c_\sigma - |\bar{E}_\sigma|} = \mathbb{1}_{\bar{T}_\sigma \text{ is a forest}} + O(N^{-1}).$$

By (43), (44) and (45), if we denote by $\delta_N(\sigma) = \delta_N(\{1, \dots, n\})$ we get

$$(46) \quad \mathbb{E}[Z_N(T_1) \dots Z_N(T_n)] = \sum_{\pi \in \mathcal{P}_2(n)} \sum_{\substack{\sigma \in \mathcal{P}_\#(V_1, \dots, V_n) \\ \text{s.t. } \bar{\sigma} = \pi}} \mathbb{1}_{\bar{T}_\sigma \text{ is a forest of } \frac{n}{2} \text{ trees}} \delta_N(\sigma) + O(N^{-1})$$

$$(47) \quad = \sum_{\pi \in \mathcal{P}_2(n)} \prod_{\{i, j\} \in \pi} \sum_{\sigma \in \mathcal{P}_\#(V_i, V_j)} \mathbb{1}_{\bar{T}_\sigma \text{ is a tree}} \delta_N(\sigma) + O(N^{-1}),$$

where we have used the independence of the entries of A to split δ_N . The case $n = 2$ gives

$$\mathbb{E}[Z_N(T_1)Z_N(T_2)] = \underbrace{\sum_{\sigma \in \mathcal{P}_\#(V_1, V_2)} \mathbb{1}_{\bar{T}_\sigma \text{ is a tree}} \delta(\sigma)}_{M^{(2)}(T_1, T_2)} + o(1),$$

where $\delta(\sigma) = \lim_{N \rightarrow \infty} \delta_N(\sigma)$, which exists since \bar{T}_σ is a tree. Indeed, in the definition (42) of $\delta(\sigma)$, we have $p_k = q_{k, \ell} = 0$ for any $k \neq \ell$. Moreover, we obtain that $\delta(\sigma) = \tau^0[T_\sigma]$ defined in Proposition 2.1, and the sum over σ on $\mathcal{P}_\#(V_1, V_2)$ can be replaced by a sum over graphs T obtained by identifying certain adjacent vertices of T_1 with adjacent edges T_2 , since $\tau^0[T]$ is zero if T is not a fat tree. We then obtain as expected the limiting covariance

$$M^{(2)}(T_1, T_2) = \sum_{T \in \mathcal{P}_\#(T_1, T_2)} \tau^0[T].$$

The general case $n \geq 3$ in (47) gives the Wick formula

$$\mathbb{E}[Z_N(T_1) \dots Z_N(T_n)] = \sum_{\pi \in \mathcal{P}_2(n)} \prod_{\{i, j\} \in \pi} M^{(2)}(T_i, T_j) + o(1),$$

which characterizes the Gaussian distribution. \square

Remark that up to (47) the errors terms are of order $O(N^{-1})$, and so if $N\mathbb{E}[a_{11}^{2k}] = C_k + o(\sqrt{N}^{-1})$ for any $k \geq 1$, then $\delta_N(B) = \delta(B) + o(\sqrt{N}^{-1})$ so that we find that

$$\mathbb{E}\left[\frac{1}{N} \text{Tr } A^k\right] = \lim_{N \rightarrow \infty} \mathbb{E}\left[\frac{1}{N} \text{Tr } A^k\right] + o(N^{-1/2})$$

and therefore we obtain the same CLT if we recenter with the limit or the expectation, as noticed in the introduction.

3. CLT FOR STIELTJES TRANSFORM AND THE METHOD OF MARTINGALES

Let $A = [a_{i,j}]$ be an $N \times N$ matrix satisfying Hypothesis 1.1. First of all, by the resolvent formula, we know that for any $z \in \mathbb{C} \setminus \mathbb{R}$, there is a constant C such that for any pair X, Y of $N \times N$ symmetric matrices,

$$|\operatorname{Tr}((z - X)^{-1} - (z - Y)^{-1})| \leq C \operatorname{rank}(X - Y).$$

Hence one can suppose that in Hypothesis 1.1, b is centered.

For any z in $\mathbb{C} \setminus \mathbb{R}$, recall that $G(z) = (z - A)^{-1}$. To prove Theorem 1.5, we show that any linear combination of the random variables

$$Z_N(z) := \frac{1}{\sqrt{N}} \operatorname{Tr} G(z) - \mathbb{E} \left[\frac{1}{\sqrt{N}} \operatorname{Tr} G(z) \right], \quad z \in \mathbb{C} \setminus \mathbb{R},$$

and their complex transposes converges in distribution to a complex Gaussian variable. Since $\overline{G(z)} = G(\bar{z})$, it is enough to fix a linear combination

$$(48) \quad M(N) := \sum_{i=1}^p \lambda_i Z_N(z_i),$$

for some fixed $p \geq 1$ and $\lambda_1, \dots, \lambda_p \in \mathbb{C}$, $z_1, \dots, z_p \in \mathbb{C} \setminus \mathbb{R}$, and prove that $M(N)$ is asymptotically Gaussian with the adequate covariance.

3.1. The method of martingales differences and reduction to the case $p = 1$. For all N , we have

$$M(N) - \mathbb{E}[M(N)] = \sum_{k=1}^N X_k \quad \text{with} \quad X_k := (\mathbb{E}_k - \mathbb{E}_{k-1})[M(N)],$$

where \mathbb{E}_k denotes the conditional expectation with respect to the σ -algebra generated by the $k \times k$ upper left corner of A . In view of Theorem 7.8 of the appendix about fluctuations of martingales, it is enough to prove that we have

$$(49) \quad \sum_{k=1}^N \mathbb{E}_{k-1}[X_k^2] \xrightarrow{N \rightarrow \infty} \sum_{i,j=1}^p \lambda_i \lambda_j C(z_i, z_j),$$

$$(50) \quad \sum_{k=1}^N \mathbb{E}_{k-1}[|X_k|^2] \xrightarrow{N \rightarrow \infty} \sum_{i,j=1}^p \lambda_i \overline{\lambda_j} C(z_i, \bar{z}_j)$$

and that for each $\varepsilon > 0$,

$$(51) \quad \sum_{k=1}^N \mathbb{E}[|X_k|^2 \mathbb{1}_{|X_k| \geq \varepsilon}] \xrightarrow{N \rightarrow \infty} 0.$$

Notice first that (49) implies (50). Let us now prove (51). The proof of (49) will then be the main difficulty of the proof of Theorem 1.5.

Proof of (51). Let A_k be the symmetric matrix with size $N-1$ obtained by removing the k -th row and the k -th column of A and set $G_k(z) := \frac{1}{z-A_k}$. Note that $\mathbb{E}_k G_k(z) = \mathbb{E}_{k-1} G_k(z)$, so that we can write

$$X_k = \sum_{i=1}^p \lambda_i \times (\mathbb{E}_k - \mathbb{E}_{k-1}) \left[\frac{1}{\sqrt{N}} (\text{Tr } G(z_i) - \text{Tr } G_k(z_i)) \right].$$

Hence by (106) of Lemma 7.5 in the appendix, there is C such that for all N and all k ,

$$|X_k| \leq \frac{C \max_i |\lambda_i| |\Im z_i|^{-1}}{\sqrt{N}}.$$

Thus for N large enough, we have that for all k , $|X_k|^2 \mathbb{1}_{|X_k| \geq \varepsilon} = 0$ and (51) is proved. \square

We now pass to the main part of the proof of Theorem 1.5, namely the proof of (49). It is divided into several steps.

We can get rid of the linear combination in (48) and assume $p = 1$. As $\overline{\text{Tr } G(z_i)} = \text{Tr } G(\bar{z}_i)$, both X_k^2 and $|X_k|^2$ are linear combinations of terms of the form

$$(\mathbb{E}_k - \mathbb{E}_{k-1}) \left[\frac{1}{\sqrt{N}} \text{Tr } G(z) \right] \times (\mathbb{E}_k - \mathbb{E}_{k-1}) \left[\frac{1}{\sqrt{N}} \text{Tr } G(z') \right],$$

with $z, z' \in \mathbb{C} \setminus \mathbb{R}$. As a consequence, we shall only fix $z \in \mathbb{C} \setminus \mathbb{R}$ and prove that for

$$Y_k(z) := (\mathbb{E}_k - \mathbb{E}_{k-1}) \left[\frac{1}{\sqrt{N}} \text{Tr } G(z) \right],$$

we have the convergence in probability for any $z, z' \in \mathbb{C} \setminus \mathbb{R}$

$$(52) \quad C_N := \sum_{k=1}^N \mathbb{E}_{k-1} [Y_k(z) Y_k(z')] \xrightarrow{N \rightarrow \infty} C(z, z').$$

First, for G_k as introduced in the proof of (51) above, as $\mathbb{E}_k G_k(z) = \mathbb{E}_{k-1} G_k(z)$ again, we have $Y_k(z) = (\mathbb{E}_k - \mathbb{E}_{k-1}) \left[\frac{1}{\sqrt{N}} (\text{Tr } G(z) - \text{Tr } G_k(z)) \right]$. Hence by Lemma 7.5, we can write

$$(53) \quad Y_k(z) = \frac{1}{\sqrt{N}} (\mathbb{E}_k - \mathbb{E}_{k-1}) \frac{1 + \mathbf{a}_k^* G_k(z)^2 \mathbf{a}_k}{z - a_{kk} - \mathbf{a}_k^* G_k(z) \mathbf{a}_k},$$

where \mathbf{a}_k is the k -th column of A where the k -th entry has been removed. To prove (52), we shall first show that we can get rid of the off diagonal terms $\sum_{j \neq \ell} \mathbf{a}_k(j) \mathbf{a}_k(\ell) G_k(z)_{j\ell}$ and $\sum_{j \neq \ell} \mathbf{a}_k(j) \mathbf{a}_k(\ell) (G_k(z)^2)_{j\ell}$ in the above expression.

3.2. Removing the off-diagonal terms. In this section, we prove that we can replace Y_k in (52) by

$$(54) \quad \tilde{Y}_k(z) := \frac{1}{\sqrt{N}}(\mathbb{E}_k - \mathbb{E}_{k-1}) \frac{1 + \sum_{j=1}^{N-1} \mathbf{a}_k(j)^2 (G_k(z)^2)_{jj}}{z - \sum_{j=1}^{N-1} \mathbf{a}_k(j)^2 G_k(z)_{jj}}.$$

Note first that by Equation (109) of Lemma 7.6 in the appendix, we have the deterministic bound, for all $z \in \mathbb{C} \setminus \mathbb{R}$,

$$\left| \frac{1 + \sum_j \mathbf{a}_k(j)^2 (G_k(z)^2)_{jj}}{z - \sum_j \mathbf{a}_k(j)^2 G_k(z)_{jj}} \right| \leq 2|\Im z|^{-1}$$

hence we deduce

$$(55) \quad |\tilde{Y}_k| \leq \frac{4|\Im z|^{-1}}{\sqrt{N}}.$$

Lemma 3.1. *Let $\tilde{Y}_k(z')$ be defined in the same way as $\tilde{Y}_k(z)$ in (54), replacing z by z' . Set also*

$$(56) \quad \tilde{C}_N := \sum_{k=1}^N \mathbb{E}_{k-1}[\tilde{Y}_k(z)\tilde{Y}_k(z')].$$

Then for C_N defined as in (52), as $N \rightarrow \infty$, we have the convergence

$$(57) \quad C_N - \tilde{C}_N \xrightarrow{L^1} 0.$$

Proof. We have

$$C_N - \tilde{C}_N = \sum_{k=1}^N \mathbb{E}_{k-1}[Y_k(z)Y_k(z') - \tilde{Y}_k(z)\tilde{Y}_k(z')] = \int_{u=0}^1 h_N(u) du,$$

with $h_N(u) := N\mathbb{E}_{k-1}[Y_k(z)Y_k(z') - \tilde{Y}_k(z)\tilde{Y}_k(z')]$ for $k = \lceil Nu \rceil$. As we already saw, by Lemmas 7.5 and 7.6 of the appendix, there is a constant C (independent of k and N) such that $|Y_k|$, $|Y'_k|$, $|\tilde{Y}_k|$ and $|\tilde{Y}_k(z')|$ are all bounded above by C/\sqrt{N} , so that $|h_N(u)| \leq 2C$. Hence by dominated convergence, it suffices to prove that for any fix u , $\|h_N(u)\|_{L^1} \rightarrow 0$, which is equivalent to $h_N(u) \rightarrow 0$ in probability. This is a direct consequence of Lemma 7.7 of the appendix (which can be applied here because we explained, at the beginning of Section 3, that one can suppose the random variable b of Hypothesis 1.1 to be centered) and of the fact that the function of two complex variables $\varphi(x_1, x_2) := \frac{1+x_1}{z-x_2}$ has a uniformly bounded gradient on the set $\{(x_1, x_2) \in \mathbb{C}^2; \Im x_2 \leq 0\}$. \square

It remains to prove that the sequence \tilde{C}_N introduced at (56) converges in probability as N goes to infinity. Note that we have

$$(58) \quad \tilde{C}_N = \int_{u=0}^1 N\mathbb{E}_{k(u)-1}[\tilde{Y}_{k(u)}(z)\tilde{Y}_{k(u)}(z')] du \quad \text{with } k(u) := \lceil Nu \rceil.$$

By (55), the integrand is uniformly bounded by $4|\Im z|^{-1}$. Hence, by dominated convergence, it is enough to prove that for any fixed $u \in (0, 1)$, as k, N tend to infinity in such a way that $k/N \rightarrow u$, we have the convergence in probability of $N\mathbb{E}_{k-1}[\tilde{Y}_k(z)\tilde{Y}_k(z')]$.

Now, set for $z, z' \in \mathbb{C} \setminus \mathbb{R}$,

$$(59) \quad f_k := \frac{1 + \sum_j \mathbf{a}_k(j)^2 (G_k(z))^2_{jj}}{z - \sum_j \mathbf{a}_k(j)^2 G_k(z)_{jj}}, \quad f'_k := \frac{1 + \sum_j \mathbf{a}_k(j)^2 (G_k(z'))^2_{jj}}{z - \sum_j \mathbf{a}_k(j)^2 G_k(z')_{jj}}.$$

We have $\sqrt{N}\tilde{Y}_k(z) = (\mathbb{E}_k - \mathbb{E}_{k-1})f_k$, so for $z, z' \in \mathbb{C} \setminus \mathbb{R}$,

$$N\mathbb{E}_{k-1}(\tilde{Y}_k(z)\tilde{Y}_k(z')) = \mathbb{E}_{k-1}[\mathbb{E}_k f_k \mathbb{E}_k f'_k] - \mathbb{E}_{k-1} f_k \mathbb{E}_{k-1} f'_k.$$

Let us denote by $\mathbb{E}_{\mathbf{a}_k}$ the expectation with respect to the randomness of the k -th column of A (i.e. the conditional expectation with respect to the σ -algebra generated by the a_{ij} 's such that $k \notin \{i, j\}$). Note that $\mathbb{E}_{k-1} = \mathbb{E}_{\mathbf{a}_k} \circ \mathbb{E}_k = \mathbb{E}_k \circ \mathbb{E}_{\mathbf{a}_k}$, hence

$$(60) \quad N\mathbb{E}_{k-1}(\tilde{Y}_k(z)\tilde{Y}_k(z')) = \mathbb{E}_{\mathbf{a}_k}[\mathbb{E}_k f_k \mathbb{E}_k f'_k] - \mathbb{E}_k \mathbb{E}_{\mathbf{a}_k} f_k \times \mathbb{E}_k \mathbb{E}_{\mathbf{a}_k} f'_k.$$

Now, we introduce (on a possibly enlarged probability space where the conditional expectations $\mathbb{E}_{k-1}, \mathbb{E}_k, \mathbb{E}_{\mathbf{a}_k}$ keep their definitions as the conditional expectations with respect to the same σ -algebras as above) an $N \times N$ random symmetric matrix

$$(61) \quad A' = [a'_{ij}]_{1 \leq i, j \leq N}$$

such that:

- the a'_{ij} 's such that $i > k$ or $j > k$ are i.i.d. copies of a_{11} (modulo the fact that A' is symmetric), independent of A ,
- for all other pairs (i, j) , $a'_{ij} = a_{ij}$,

then we have

$$\mathbb{E}_k f_k \mathbb{E}_k f'_k = \mathbb{E}_k(f_k \times f''_k),$$

where f''_k is defined out of A' in the same way as f'_k is defined out of A in (59) (note that the k -th column is the same in A and in A'). It follows that

$$(62) \quad \begin{aligned} N\mathbb{E}_{k-1}(\tilde{Y}_k(z)\tilde{Y}_k(z')) &= \mathbb{E}_{\mathbf{a}_k}[\mathbb{E}_k(f_k \times f''_k)] - \mathbb{E}_k \mathbb{E}_{\mathbf{a}_k} f_k \times \mathbb{E}_k \mathbb{E}_{\mathbf{a}_k} f'_k \\ &= \mathbb{E}_k[\mathbb{E}_{\mathbf{a}_k}(f_k \times f''_k)] - \mathbb{E}_k \mathbb{E}_{\mathbf{a}_k} f_k \times \mathbb{E}_k \mathbb{E}_{\mathbf{a}_k} f'_k. \end{aligned}$$

We shall in the sequel prove that as N tends to infinity, regardless to the value of k , we have the almost sure convergences

$$(63) \quad \mathbb{E}_{\mathbf{a}_k} f_k \rightarrow L(z) \quad \mathbb{E}_{\mathbf{a}_k} f'_k \rightarrow L(z')$$

and for any $u \in (0, 1)$, as $N, k \rightarrow \infty$ so that $k/N \rightarrow u$, we have

$$(64) \quad \mathbb{E}_{\mathbf{a}_k}(f_k \times f''_k) \rightarrow \Psi^u(z, z') \text{ a.s.},$$

where $L(z, z') = \int_0^1 \Psi^u(z, z') du$. The convergences of (63) and (64) are based on an abstract convergence result stated in next section, where we use the convergence of generalized moments of Proposition 2.1. They are stated in Lemmas 3.4 and 3.5 respectively.

Note that by Lemma 7.6,

$$(65) \quad |f_k| \leq 4|\Im z|^{-1},$$

so once (63) and (64) proved, we will have the convergence in L^2

$$\mathbb{E}_k[\mathbb{E}_{\mathbf{a}_k}(f_{k(u)} \times f''_{k(u)})] - \mathbb{E}_k \mathbb{E}_{\mathbf{a}_k} f_k \times \mathbb{E}_k \mathbb{E}_{\mathbf{a}_k} f'_k \longrightarrow \Psi^u(z, z') - L(z)L(z').$$

Thus by (62), we will have proved the convergence of (60), hence completing the proof of the theorem.

3.3. An abstract convergence result. Remember that $A_k = (a_{ij})$ is the square matrix of size $N - 1$ obtained by removing the k -th row and the k -th column of A , that $G_k(z) = (z - A_k)^{-1}$ and that A'_k is a copy of A_k where the entries (i, j) for i or $j \geq k$ are independent of A_k and the other are those of A_k . We denote $G'_k(z') = (z' - A'_k)^{-1}$.

Lemma 3.2. *Under Hypothesis 1.1 and condition (14), as N goes to infinity and $\frac{k}{N}$ tends to u in $(0, 1)$, the random probability measure on \mathbb{C}^2*

$$(66) \quad \nu_{z, z'}^{k, N} := \frac{1}{N-1} \sum_{j=1}^{N-1} \delta_{\{G_k(z)_{jj}, G'_k(z')_{jj}\}}, \quad z, z' \in \mathbb{C} \setminus \mathbb{R}.$$

converges weakly almost surely to a deterministic probability measure $\nu_{z, z'}^u$ on \mathbb{C}^2 .

Proof. By e.g. Theorem C.8 of [1], it is enough to prove that for any bounded and Lipschitz function f , $\nu_{z, z'}^{k, N}(f)$ converges almost surely to $\nu_{z, z'}^u(f)$. Moreover, adapting the proof of Lemma 7.4, one can easily see that for any such f ,

$$\nu_{z, z'}^{k, N}(f) - \mathbb{E}[\nu_{z, z'}^{k, N}(f)]$$

converges almost surely to zero. The only modification of the proof is to complete the resolvent identity by noticing that

$$(z - A_k)^{-2} - (z - A_k^B)^{-2} = (z - A_k^B)^{-2} (z(A_k - A_k^B) - (A_k - A_k^B)^2 - 2A_k^B(A_k - A_k^B))(z - A_k)^{-2},$$

which gives that this matrix has rank bounded by $3 \times \text{rank}(A - B)$. Hence it is enough to prove that the deterministic sequence of measures $\mathbb{E}[\nu_{z, z'}^{k, N}(\cdot)]$ converges weakly (the measure $\mathbb{E}[\nu_{z, z'}^{k, N}(\cdot)]$ is defined by $\mathbb{E}[\nu_{z, z'}^{k, N}(\cdot)](f) = \mathbb{E}[\nu_{z, z'}^{k, N}(f)]$ for any bounded Borel function f : one can easily verify that this is actually a probability measure). Moreover, the measure $\mathbb{E}[\nu_{z, z'}^{k, N}(\cdot)]$ always belongs to the set of probability measures supported by the compact subset $B(0, |\Im z|^{-1})^2$. The set of such probability measures is compact for the topology of weak convergence. Hence it suffices to prove that it admits at most one accumulation point.

Let us first explain how the $a = b_\varepsilon + c_\varepsilon$ decomposition of Hypothesis 1.1 allows to reduce the problem to the case of Wigner matrices with exploding moments. For any $\varepsilon > 0$, the matrix $A^\varepsilon := [b_{\varepsilon, ij} - \mathbb{E}b_{\varepsilon, ij}]_{i, j=1}^N$ is a Wigner matrix with exploding moments by Hypothesis 1.1.

Claim : Almost surely, $\limsup_{N \rightarrow \infty} \frac{\text{rank}(A - A^\varepsilon)}{N} \leq \varepsilon$.

Indeed, the rank of $A - A^\varepsilon$ is at most $1 + \sum_{i=1}^N \mathbb{1}_{\exists j \leq i, c_{\varepsilon,ij} \neq 0}$. Hence, by independence and concentration inequalities such as Azuma-Hoeffding inequality, it suffices to prove that

$$\limsup_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N \mathbb{P}(\exists j \leq i, c_{\varepsilon,ij} \neq 0) \leq \frac{\varepsilon}{2}.$$

But we have, for $p := \mathbb{P}(c_\varepsilon \neq 0)$,

$$\frac{1}{N} \sum_{i=1}^N \mathbb{P}(\exists j \leq i, c_{\varepsilon,ij} \neq 0) = \frac{1}{N} \sum_{i=1}^N 1 - (1-p)^i = 1 - (1-p) \frac{1 - (1-p)^N}{pN}$$

As pN is small, $(1-p)^N = 1 - Np + O(Np^2)$,

$$\frac{1}{N} \sum_{i=1}^N \mathbb{P}(\exists j \leq i, c_{\varepsilon,ij} \neq 0) = O(Np)$$

and the Claim is proved, by Hypothesis (5).

We can adapt arguments of [10] (see also Equation (91) of [14]) to see that it is enough to prove the weak convergence of $\mathbb{E}[\nu_{z,z'}^{k,N}(\cdot)]$ when A is a Wigner matrix with exploding moments. Since these measures are uniformly compactly supported, it is sufficient to prove the convergence in moments.

We consider a polynomial $P = x_1^{n_1} \overline{x_1}^{m_1} x_2^{n_2} \overline{x_2}^{m_2}$ and remark that

$$\begin{aligned} \mathbb{E}[\nu_{z,z'}^{k,N}(P)] &= \mathbb{E} \left[\frac{1}{N-1} \sum_{j=1}^{N-1} (G_k(z)_{jj})^{n_1} (G_k(\bar{z})_{jj})^{m_1} (G'_k(z')_{jj})^{n_2} (G'_k(\bar{z}')_{jj})^{m_2} \right] \\ &= \mathbb{E} \left[\frac{1}{N-1} \text{Tr} \left[G_k(z)^{\circ n_1} \circ G_k(\bar{z})^{\circ m_1} \circ G'_k(z')^{\circ n_2} \circ G'_k(\bar{z}')^{\circ m_2} \right] \right], \end{aligned}$$

where \circ denotes the Hadamard (entry-wise) product of matrices and

$$M^{\circ n} := \underbrace{M \circ \cdots \circ M}_n.$$

We set $\ell_1 = n_1 + m_1$ and $\ell_2 = n_2 + m_2$. Let $(Y_i, Y'_j)_{i=1, \dots, \ell_1, j=1, \dots, \ell_2}$ be a family of random variables such that for any $p_i, q_j \geq 0$,

$$(67) \quad \mathbb{E} \left[\prod_{i=1}^{\ell_1} Y_i^{p_i} \prod_{j=1}^{\ell_2} Y'_j{}^{q_j} \right] = \mathbb{E} \left[\frac{1}{N-1} \text{Tr} \left[A_k^{p_1} \circ \cdots \circ A_k^{p_{\ell_1}} \circ A'_k{}^{q_1} \circ \cdots \circ A'_k{}^{q_{\ell_2}} \right] \right].$$

Such a family exists by Proposition 7.10. By [38, Proposition 3.10], the couple of random matrices (A, A') satisfies the so-called convergence in distribution of traffics, so that the RHS converges. For the reader's convenience, we give the limiting value of (67), even if we

do not use it later. It is obtained by applying the rule of the so-called traffic freeness in [38].

$$(68) \quad \mathbb{E} \left[\prod_{i=1}^{\ell_1} Y_i^{p_i} \prod_{j=1}^{\ell_2} Y_j'^{q_j} \right] \xrightarrow{N \rightarrow \infty} \sum_{\pi \in \mathcal{P}(V_T)} \tau^0[T^\pi] \alpha(T^\pi),$$

where

1. we have considered T the graph whose edges are labelled by indeterminates a and a' , obtained by
 - considering the disjoint union of the graphs with vertices $1, \dots, p_i$ and edges $\{1, 2\}, \dots, \{p_i - 1, p_i\}, \{p_i, 1\}$ labelled a , $i = 1, \dots, \ell_1$,
 - considering the disjoint union of the graphs with vertices $1, \dots, q_j$ and edges $\{1, 2\}, \dots, \{q_j - 1, q_j\}, \{q_j, 1\}$ labelled a' , $j = 1, \dots, \ell_2$,
 - identifying the vertex 1 of each of these graphs (we get a connected graph, bouquet of cycles),
2. we have denoted by V_T the set of vertices of T , $\mathcal{P}(V_T)$ is the set of partitions of V_T and T^π denotes the graph obtained by identifying the vertices of T that belong to a same block of π ,
3. the quantity $\tau^0[T^\pi]$ is as in Proposition 2.1,
4. we have set

$$\alpha(T^\pi) = \sum_{V_\pi = V_1 \sqcup V_2} \mathbb{1}_{\left\{ \begin{array}{c} \text{the edges which are not linking} \\ \text{adjacent vertices of } V_1 \text{ have the same label} \end{array} \right\}} u^{|V_1|} (1 - u)^{|V_2|},$$

where the sum is over all partitions of the set V_π of vertices of T^π and $|E_i|$ is the number of edges between adjacent vertices of V_i , $i = 1, 2$.

Formula (68) could also be derived by the same techniques than those developed in Section 2. The random variables Y_i and Y_j' are distributed according to the limiting eigenvalues distribution of A . Recall that we assume that the sequence $(C_k)_{k \geq 2}$ defined in (9) satisfies $C_k \leq C^k$ for a constant $C > 0$. Then, following the proof of [49, Proposition 10], the exponential power series of the limiting eigenvalues distribution of A has a positive radius of convergence. So, by a generalization of [12, Theorem 30.1] to the multi-dimensional case, we get that the distribution of (Y_i, Y_j') is characterized by its moments. Then, we get that (Y_i, Y_j') converges weakly to a family of random variables (y_i, y_j') . We set $f_z(y) = (z - y)^{-1}$. We then obtain the convergence

$$\mathbb{E}[\nu_{z, z'}^{k, N}(P)] = \mathbb{E} \left[\prod_{i=1}^{n_1} f_z(Y_i) \prod_{j=1}^{n_2} f_{z'}(Y_j') \right] \xrightarrow{N \rightarrow \infty} \mathbb{E} \left[\prod_{i=1}^{n_1} f_z(y_i) \prod_{j=1}^{n_2} f_{z'}(y_j') \right]$$

Hence $\mathbb{E}[\nu_{z, z'}^{k, N}(\cdot)]$ converges weakly. □

This convergence could also be proven without Proposition 7.10 but with appropriate bounds on the growth of moments. We have the following Corollary.

Corollary 3.3. (1) For $z \in \mathbb{C} \setminus \mathbb{R}$, t so that $\Im z t \geq 0$,

$$\rho_z^N(t) := \frac{1}{N-1} \sum_{i=1}^{N-1} \Phi(tG_k(z)_{ii}) \xrightarrow{N \rightarrow \infty} \rho_z(t) := \int \Phi(tx) \nu_{z,z'}^0(dx, dx') \quad a.s.$$

(2) For $z, z' \in \mathbb{C} \setminus \mathbb{R}$, t, t' so that $\Lambda = \{t\Im z \geq 0, t'\Im z' \geq 0, |t| + |t'| > 0\}$ and k/N going to u ,

$$\rho_{z,z'}^{N,k}(t, t') := \frac{1}{N-1} \sum_{i=1}^{N-1} \Phi(tG_k(z)_{ii} + t'G'_k(z')_{ii})$$

converges almost surely towards

$$(69) \quad \rho_{z,z'}^u(t, t') = \int \Phi(tx + t'x') \nu_{z,z'}^u(dx, dx').$$

(3) For any $(t, t') \in \mathbb{R}$, the functions $\rho^{N,k}(t, t')$ are analytic on $\Lambda = \{z, z' : t\Im z > 0, t'\Im z' > 0\}$ and uniformly bounded on compact subsets of $\bar{\Lambda}$. They have a non positive real part. Their limits are also analytic on Λ and have a non positive real part.

(4) For all $z \in \mathbb{C}^+$,

$$\lim_{N \rightarrow \infty} \frac{1}{N} \text{Tr}((z - A)^{-1}) = i \int_0^\infty e^{itz + \rho_z(t)} dt \quad a.s.$$

Proof. Let us first notice that $\rho_z^N(t) = \rho_{z,z'}^{N,k}(t, 0)$ for all k so we only need to focus on $\rho_{z,z'}^{N,k}(t, t')$. The point wise convergence of the function $\rho^{N,k}$ is a direct consequence of the continuity of Φ (recall we assumed that Φ extends continuously to the real line), of the boundedness of $tG_k(z)_{ii} + t'G'_k(z')_{ii}$ (by $t/\Im z + t'/\Im z'$) and Lemma 3.2. To show analyticity, note that for all $j \in \{1, \dots, N\}$, $G(z)_{jj}$ is an analytic function on \mathbb{C}^+ , taking its values in \mathbb{C}^- (and vice versa) almost surely by (107). Hence, on Λ , $tG_k(z)_{ii} + t'G'_k(z')_{ii}$ is an analytic function with values in \mathbb{C}^- . Therefore, as Φ is analytic on \mathbb{C}^- , $\rho^{N,k}$ is an analytic function on Λ almost surely. Moreover, as Φ extends continuously to the real line, it is uniformly bounded on compact subsets of $\overline{\mathbb{C}^-}$ and hence $\rho^{N,k}$ is uniformly bounded on compact subsets of Λ . This implies by Montel's theorem that the limit ρ^u of $\rho^{N,k}$ is also analytic. Finally, $\rho^{N,k}$ has non positive real part as the image of \mathbb{C}^- by Φ . Indeed, as $\Re(-i\lambda|a_{11}|^2) \leq 0$, we have

$$(70) \quad |\phi_N(\lambda)| = |\mathbb{E}(e^{-i\lambda|a_{11}|^2})| \leq 1 \quad \Rightarrow \quad \Re \Phi(\lambda) = \lim_{N \rightarrow \infty} N \log |\phi_N(\lambda)| \leq 0.$$

For the last point, first note that by Lemma 7.3, it is enough to prove the result for $\mathbb{E}[N^{-1} \text{Tr} G(z)]$ instead of $N^{-1} \text{Tr} G(z)$. Second, by exchangeability, $\mathbb{E}[N^{-1} \text{Tr} G(z)] = \mathbb{E}[G(z)_{11}]$. Remind that \mathbf{a}_k is the k -th column of A (or A') where the k -th entry has been removed. Using Schur complement formula and getting rid of the off diagonal terms (by arguments similar to those of Section 3.2, *i.e.* essentially Lemma 7.7), we have for any

$z \in \mathbb{C}^+$

$$\begin{aligned}\mathbb{E}[G(z)_{11}] &= \mathbb{E}\left[\frac{1}{z - \sum_{j=1}^{N-1} |\mathbf{a}_1(j)|^2 G_1(z)_{jj} + \varepsilon_N}\right] \\ &= \mathbb{E}\left[\frac{1}{z - \sum_{j=1}^{N-1} \mathbf{a}_1(j)^2 G_1(z)_{jj}}\right] + o(1).\end{aligned}$$

Remember that we have for $\lambda \in \mathbb{C} \setminus \mathbb{R}$,

$$(71) \quad \frac{1}{\lambda} = -i \int_0^{\operatorname{sgn} \lambda \infty} e^{it\lambda} dt,$$

and that $\phi_N(\lambda) = \mathbb{E}[\exp(-i\lambda|a_{11}|^2)]$, which gives

$$\begin{aligned}\mathbb{E}[G(z)_{11}] &= i \int_0^\infty \mathbb{E}[e^{i\lambda(z - \sum_{j=1}^{N-1} \mathbf{a}_1(j)^2 G_1(z)_{jj})}] d\lambda + o(1) \\ &= i \int_0^\infty e^{i\lambda z} \mathbb{E}\left[\prod_{j=1}^{N-1} \phi_N(\lambda G_1(z)_{jj})\right] d\lambda + o(1).\end{aligned}$$

We used the exponential decay to switch the integral and the expectation. This also allows to truncate the integral: for any $M \geq 0$,

$$\int_0^\infty e^{i\lambda z} \mathbb{E}\left[\prod_{j=1}^{N-1} \phi_N(\lambda G_1(z)_{jj})\right] d\lambda = \int_0^M e^{i\lambda z} \mathbb{E}\left[\prod_{j=1}^{N-1} \phi_N(\lambda G_1(z)_{jj})\right] d\lambda + \epsilon_{m,M,z,N},$$

where $\epsilon_{m,M,z,N}$ goes to zero as M tends to infinity, uniformly on N and on the randomness. Remember that by assumption (7), we have

$$(72) \quad N(\phi_N(\lambda) - 1) \xrightarrow{N \rightarrow \infty} \Phi(\lambda), \quad \forall \lambda \in \overline{\mathbb{C}}^-$$

where the convergence is uniform on compact subsets of $\overline{\mathbb{C}}^-$. Hence, since for $|\Im z| \geq \delta > 0$, $|t| \leq M$, $tG_k(z)_{jj}$ belongs to the compact set $\{\lambda \in \overline{\mathbb{C}}^-; |\lambda| \leq M/\delta\}$, we have

$$\prod_{j=1}^{N-1} \phi_N(tG_k(z)_{jj}) = \exp\left(\frac{1}{N} \sum_{j=1}^{N-1} \Phi(tG_k(z)_{jj})\right) + \varepsilon_{t,z,N}^{(1)} = \exp(\rho_z^N(t)) + \varepsilon_{t,z,N}^{(1)}$$

where $\varepsilon_{t,z,N}^{(1)}$ converges almost surely to zero as N goes to infinity. By Corollary 3.3,

$$\exp\left(\frac{1}{N} \sum_{j=1}^{N-1} \Phi(tG_k(z)_{jj})\right) = \exp(\rho_z(t)) + \varepsilon_{t,z,N}^{(2)},$$

where $\varepsilon_{t,z,N}^{(2)}$ converges to zero almost surely. Hence, we deduce the almost sure convergence

$$(73) \quad e^{itz} \prod_{j=1}^{N-1} \phi_N(tG_k(z)_{jj}) \xrightarrow{N \rightarrow \infty} e^{itz + \rho_z(t)}.$$

As ρ_z^N has non positive real part, we conclude by dominated convergence theorem and by getting rid of the truncation of the integral. \square

3.4. Computation and convergence of $\mathbb{E}_{\mathbf{a}_k} f_k$ and $\mathbb{E}_{\mathbf{a}_k} f'_k$.

Lemma 3.4. *Almost surely, we have the convergence*

$$(74) \quad \mathbb{E}_{\mathbf{a}_k} f_k \xrightarrow{N \rightarrow \infty} L(z) := - \int_0^{\operatorname{sgn}_z \infty} \frac{1}{t} \partial_z e^{itz + \rho_z(t)} dt.$$

where $\operatorname{sgn}_z := \operatorname{sgn}(\Im z)$ and $\rho_z(t)$ is defined in Corollary 3.3.

Proof. Remember that for any $z \in \mathbb{C} \setminus \mathbb{R}$,

$$f_k = \frac{1 + \sum_j \mathbf{a}_k(j)^2 (G_k(z)^2)_{jj}}{z - \sum_j \mathbf{a}_k(j)^2 G_k(z)_{jj}} := \frac{\partial_z \lambda_N(z)}{\lambda_N(z)}.$$

Hence, by (71), since by (107) the sign of the imaginary part of $\lambda_N(z)$ is sgn_z , the random variable f_k can be written

$$(75) \quad \begin{aligned} f_k &= -i \int_0^{\operatorname{sgn}_z \infty} \partial_z \lambda_N(z) e^{it\lambda_N(z)} dt. \\ &= -i \int_{\operatorname{sgn}_z m}^{\operatorname{sgn}_z \infty} \partial_z \lambda_N(z) e^{it\lambda_N(z)} dt - i \int_0^{\operatorname{sgn}_z m} \partial_z \lambda_N(z) e^{it\lambda_N(z)} dt \\ &:= \tilde{f}_{k,m} + \eta_m(z), \end{aligned}$$

where $m > 0$ and $\eta_m(z) = \frac{\partial_z \lambda_N(z)}{\lambda_N(z)} (1 - e^{i \operatorname{sgn}_z m \lambda_N(z)})$.

We next show that for all $\varepsilon > 0$ there exists m_0 so that for $m < m_0$, N large enough

$$(76) \quad \mathbb{E}_{\mathbf{a}_k} [|\eta_m(z)|] \leq \varepsilon$$

By (109) and since the sign of the imaginary part of $\lambda_N(z)$ is sgn_z , one has $|\eta_m(z)| \leq 4|\Im z|^{-1}$. More precisely, for any $K > 0$, we find

$$|\eta_m(z)| \leq \frac{4}{\Im z} (mK |\Im z|^{-1} + \mathbb{1}_{\sum_j \mathbf{a}_k(j)^2 \geq K}).$$

By (4), we deduce that for any $\varepsilon > 0$ we can choose m small enough so that for N large enough

$$(77) \quad \mathbb{E}_{\mathbf{a}_k} [|\eta_m(z)|] \leq C\varepsilon.$$

In the following we shall therefore neglect the term $\eta_m(z)$. By (107) and (108) we have the following bound: for any $t \neq 0$ and any $z \in \mathbb{C} \setminus \mathbb{R}$ such that $|\Im z| \geq \delta > 0$,

$$(78) \quad \left| \sum_{j=1}^{N-1} \mathbf{a}_k(j)^2 (G_k(z)^2)_{jj} e^{-it \sum \mathbf{a}_k(j)^2 G_k(z)_{jj}} \right| \leq \frac{1}{\delta} \sup_{x \geq 0} x e^{-tx} = \frac{e^{-1}}{t\delta}$$

so that for M large enough and $|\Im z| \geq \delta > 0$,

$$\mathbb{E}_{\mathbf{a}_k} \tilde{f}_{k,m} = -i \int_{\operatorname{sgn}_z m}^{\operatorname{sgn}_z M} \mathbb{E}_{\mathbf{a}_k} [\partial_z \lambda_N(z) e^{it\lambda_N(z)}] dt + \varepsilon_{m,M,z,N}$$

where $\varepsilon_{m,M,z,N}$ is arbitrary small as M is large, uniformly on N and on the randomness. Moreover, by (78) one has

$$(79) \quad \mathbb{E}_{\mathbf{a}_k} [\partial_z \lambda_N(z) e^{it\lambda_N(z)}] = \frac{1}{it} \partial_z (\mathbb{E}_{\mathbf{a}_k} [e^{it\lambda_N(z)}])$$

Recall that $\phi_N(\lambda) = \mathbb{E}[\exp(-i\lambda|a_{11}|^2)]$, so we have

$$(80) \quad \mathbb{E}_{\mathbf{a}_k} [e^{it\lambda_N(z)}] = \mathbb{E}_{\mathbf{a}_k} \left[e^{itz - it \sum_j \mathbf{a}_k(j)^2 G_k(z)_{jj}} \right] = e^{itz} \prod_{j=1}^{N-1} \phi_N(tG_k(z)_{jj}).$$

Remind that in (73) we have shown in the proof of the last item of Corollary 3.3

$$(81) \quad e^{itz} \prod_{j=1}^{N-1} \phi_N(tG_k(z)_{jj}) \xrightarrow[N \rightarrow \infty]{} e^{itz + \rho_z(t)} \quad a.s.$$

As in the proof of Corollary 3.3, since the left hand side is analytic and uniformly bounded, we deduce by Montel's theorem that its convergence entails the convergence of its derivatives. We then get by (80), for all t, z so that $t/\Im z > 0$, the almost sure convergence

$$\partial_z \left(\mathbb{E}_{\mathbf{a}_k} \left[e^{itz - it \sum_j \mathbf{a}_k(j)^2 G_k(z)_{jj}} \right] \right) \xrightarrow[N \rightarrow \infty]{} \partial_z (e^{itz + \rho_z(t)}).$$

We then obtain by dominated convergence (remember the integrant is uniformly bounded by (79) and (78)),

$$\mathbb{E}_{\mathbf{a}_k} \tilde{f}_{k,m} = - \int_{\operatorname{sgn}_z m}^{\operatorname{sgn}_z M} \frac{1}{t} \partial_z e^{itz + \rho_z(t)} dt + \tilde{\varepsilon}_{m,M,z,N},$$

where $\tilde{\varepsilon}_{m,M,z,N}$ converges to zero almost surely as N goes to infinity.

By (79) and estimate (78), we have the estimate

$$\left| \frac{1}{t} \partial_z e^{itz + \rho_z(t)} \right| \leq \left(1 + \frac{e^{-1}}{t\Im z} \right) e^{-t\Im z}$$

so we can let M going to infinity to obtain the almost sure convergence

$$\mathbb{E}_{\mathbf{a}_k} \tilde{f}_{m,k} \xrightarrow[N \rightarrow \infty]{} - \int_{\operatorname{sgn}_z m}^{\operatorname{sgn}_z \infty} \frac{1}{t} \partial_z e^{itz + \rho_z(t)} dt.$$

In the Lévy case, one has $\rho_z(t) = t^{\frac{\alpha}{2}}\rho_z(1)$, and in the exploding moments case, $|\frac{1}{t}\partial_z e^{itz+\rho_z(t)}| \leq 1 + \frac{C_2}{\Im z^2}$, so that the integral converges at zero and we obtain

$$\mathbb{E}_{\mathbf{a}_k} f_k \xrightarrow{N \rightarrow \infty} - \int_0^{\operatorname{sgn}_z \infty} \frac{1}{t} \partial_z e^{itz+\rho_z(t)} dt.$$

□

Of course, this convergence is uniform in k since the law of $G_k(z)$ does not depend on k and an analogous formula is true for $\mathbb{E}_{\mathbf{a}_k} f'_k$, replacing z by z' .

3.5. Computation of $\mathbb{E}_{\mathbf{a}_k}(f_k \times f''_k)$.

Lemma 3.5. *Almost surely, we have the convergence*

$$(82) \quad \mathbb{E}_{\mathbf{a}_k} f_k f''_k \xrightarrow{N \rightarrow \infty} \Psi^u(z, z') := \int_0^{\operatorname{sgn}_z \infty} \int_0^{\operatorname{sgn}_{z'} \infty} \frac{1}{tt'} \partial_{z,z'}^2 e^{itz+it'z'+\rho_{z,z'}^u(t,t')} dt dt',$$

where $\rho_{z,z'}^u$ is defined in Corollary 3.3.

Proof. We shall start again by using Formula (75) for f_k , and its analogue for

$$f''_k = \frac{1 + \sum_j \mathbf{a}_k(j)^2 (G'_k(z'))_{jj}}{z - \sum_j \mathbf{a}_k(j)^2 G'_k(z')_{jj}} := \frac{\partial_z \lambda'_N(z')}{\lambda'_N(z')},$$

where $G'_k(z')$ is defined as $G_k(z)$, replacing z by z' and the matrix A by the matrix A' defined by (61), which gives

$$(83) \quad \begin{aligned} f''_k &= -i \int_0^{\operatorname{sgn}_{z'} \infty} \partial_z \lambda'_N(z') e^{it' \lambda'_N(z')} dt' \\ &= -i \int_{\operatorname{sgn}'_z m}^{\operatorname{sgn}'_z \infty} \partial_z \lambda'_N(z') e^{it' \lambda'_N(z')} dt' - i \int_0^{\operatorname{sgn}_z m} \partial_z \lambda'_N(z') e^{it' \lambda'_N(z')} dt' \end{aligned}$$

The upper bound (78) allows us to bound the first term uniformly by $\log m^{-1}$ and to truncate the integrals for $\operatorname{sgn}_z t, \operatorname{sgn}_{z'} t' \leq M$. Therefore, up to a small error ε uniform for M large, m small and provided $|\Im z|, |\Im z'| \geq \delta > 0$, we have

$$(84) \quad \mathbb{E}_{\mathbf{a}_k} \tilde{f}_k \times \tilde{f}''_k = - \int_{\operatorname{sgn}_z m}^{\operatorname{sgn}_z M} \int_{\operatorname{sgn}'_z m}^{\operatorname{sgn}'_z M} \mathbb{E}_{\mathbf{a}_k} \partial_z \lambda_N(z) \partial_z \lambda'_N(z') e^{it \lambda_N(z) + it' \lambda'_N(z')} dt dt' + \varepsilon$$

As in the previous case, the upper bound (78) allows us to write

$$(85) \quad \mathbb{E}_{\mathbf{a}_k} \left[\partial_z \lambda_N(z) \partial_z \lambda'_N(z') e^{it \lambda_N(z) + it' \lambda'_N(z')} \right] = - \frac{1}{tt'} \partial_{z,z'}^2 \left(\mathbb{E}_{\mathbf{a}_k} \left[e^{it \lambda_N(z) + it' \lambda'_N(z')} \right] \right)$$

Remember that $\phi_N(\lambda) = \mathbb{E}[\exp(-i\lambda|a_{11}|^2)]$, so we have

$$(86) \quad \mathbb{E}_{\mathbf{a}_k} \left[e^{it \lambda_N(z) + it' \lambda'_N(z')} \right] = e^{itz+it'z'} \prod_{j=1}^{N-1} \phi_N(t G_k(z)_{jj} + t' G'_k(z')_{jj}).$$

By assumption (7) and Corollary 3.3, we have the following almost sure convergence as N goes to infinity and $\frac{k}{N}$ goes to u in $(0, 1)$

$$e^{itz+it'z'} \prod_{j=1}^{N-1} \phi_N(tG_k(z)_{jj} + t'G'_k(z')_{jj}) \xrightarrow{N \rightarrow \infty} e^{itz+it'z'+\rho_{z,z'}^u(t,t')}.$$

Almost surely, for any t, t' , the map $(z, z') \mapsto e^{itz+it'z'} \prod_{j=1}^{N-1} \phi_N(tG_k(z)_{jj} + t'G'_k(z')_{jj})$ is analytic on $\mathbb{C}^{\text{sgn } t} \times \mathbb{C}^{\text{sgn } t'}$ and bounded by one. Hence, with the same arguments as in the previous section, we get almost surely and for any t, t' so that $t/\Im z$ and $t'/\Im z'$ are positive, the uniform convergence for the second derivatives on compact subsets. The truncations of the integrals can be suppressed as in the previous section, we obtain the almost sure convergence

$$\mathbb{E}_{\mathbf{a}_k} f_k \times f_k'' \xrightarrow{N \rightarrow \infty} \int_0^{\text{sgn } z} \int_0^{\text{sgn } z'} \frac{1}{tt'} \partial_{z,z'}^2 e^{itz+it'z'+\rho_{z,z'}^u(t,t')} dt dt'.$$

□

Hence we have proved the convergences (63) and (64). This completes the proof of Theorem 1.5.

4. PROOF OF COROLLARY 1.6

Note first that by linearity of the map $f \mapsto Z_N(f)$, to prove the convergence in law of the process, it suffices to prove the convergence in law of $Z_N(f)$ for any fixed f . Then, by Lemma 7.9 and the concentration inequality (103) of the appendix, it suffices to prove the result for $f \in \mathcal{C}_c^2(\mathbb{R})$. For $f \in \mathcal{C}_c^2(\mathbb{R})$, we use [1, Eq. (5.5.11)] to see that for any probability measure μ

$$\int f(x) d\mu(x) = \Re \left(\int dx \int dy \bar{\partial} \Psi_f(x, y) \int \frac{1}{t - x - iy} d\mu(t) \right)$$

Applying this to the empirical measure of eigenvalues and its expectation, we deduce that

$$Z_N(f) = \Re \left(\int dx \int dy \bar{\partial} \Psi_f(x, y) Z_N(x + iy) \right).$$

Note here that $\bar{\partial} \Psi_f(x, y)$ is supported in a compact set $[-c_0, c_0] \times [0, c'_0]$ and is bounded by $c|y|$ for a finite constant c . Hence, the integral is well converging. We next show that we can approximate it by

$$Z_N^\Delta(f) = \Re \left(\int dx \int dy \bar{\partial} \Psi_f(x, y) Z_N(\Delta(x) + i\Delta(y)) \right),$$

where $\Delta(x) = \lceil 2^k x \rceil 2^{-k}$. The random variable $Z_N^\Delta(f)$ is only a finite sum of $Z_N(x + iy)$ and therefore it converges in law by Theorem 1.5, towards

$$Z^\Delta(f) = \Re \left(\int dx \int dy \bar{\partial} \Psi_f(x, y) Z(\Delta(x) + i\Delta(y)) \right),$$

We finally show the convergence in probability of $Z_N^\Delta(f) - Z_N(f)$ and $Z^\Delta(f) - Z(f)$ to zero by bounding their L^1 norms

$$\mathbb{E}[|Z_N^\Delta(f) - Z_N(f)|] \leq \int_{-c_0}^{c_0} dx \int_0^{c'_0} dy c|y| \mathbb{E}[|Z_N(\Delta(x) + i\Delta(y)) - Z_N(x + iy)|].$$

But, by Lemma 7.3, there exists a finite constant C such that for any $\varepsilon > 0$

$$\begin{aligned} \mathbb{E}[|Z_N(\Delta(x) + i\Delta(y)) - Z_N(x + iy)|] &\leq C \left\| \frac{1}{\cdot - x - iy} - \frac{1}{\cdot - \Delta(x) - i\Delta(y)} \right\|_{\text{TV}} \\ &\leq C \frac{1}{y} \left(\frac{|x - \Delta(x)| + |y - \Delta(y)|}{y} \right)^\varepsilon. \end{aligned}$$

Taking $\varepsilon \in (0, 1)$ we deduce that there exists a finite constant C_ε such that

$$\mathbb{E}[|Z_N^\Delta(f) - Z_N(f)|] \leq C_\varepsilon 2^{-\varepsilon k}.$$

Of course the same estimate holds for $\mathbb{E}[|Z^\Delta(f) - Z(f)|]$. This implies the desired convergence in probability of $|Z_N^\Delta(f) - Z_N(f)|$ and $|Z^\Delta(f) - Z(f)|$ to zero, and therefore the desired convergence in law of $Z_N(f)$. This proves the corollary.

5. FIXED POINT CHARACTERIZATIONS

In this section, we provide characterizations of the functions ρ_z and $\rho_{z,z'}^u$ involved in the covariance of the limiting process of Theorem 1.5 as fixed points of certain functions. In fact, we also give an independent proof of the existence of such limits, and of Corollary 3.3.

5.1. Fixed point characterization of $\rho_z(\cdot)$: proof of Theorem 1.9. We now prove the fixed point equation for the non random function involved in the Lemma 3.4, given for $z \in \mathbb{C}^+$ and $\lambda > 0$ by,

$$\rho_z(\lambda) = \lim_{N \rightarrow \infty} \rho_z^N(\lambda) = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{j=1}^N \Phi(\lambda G_k(z)_{jj}),$$

where we have proved that this convergence holds almost surely in Corollary 3.3. Note however that under the assumptions of Theorem 1.9, the arguments below provide another proof of this convergence where we do not have to assume (14).

We denote in short A for A_k , G for G_k and \mathbf{a} for \mathbf{a}_k in the following, and we do not detail the steps of the proof, which are very similar to those in [10] and Corollary 3.3, but outline them. Since we have already seen that ρ is analytic in Corollary 3.3 we need only to prove the fixed point equation. Let A_1 be the $(N-2) \times (N-2)$ principal submatrix of A obtained by removing the first row and the first column of A , and let $G_1(z) := (z - A_1)^{-1}$. Let \mathbf{a}_1 be the first column of A where the first entry has been removed. Using first the concentration lemma 7.4, then exchangeability of the $G(z)_{jj}$'s, then Schur complement formula (see [1, Lem. 2.4.6]), and then the fact that we can get rid of the off diagonal terms by the same

argument as in the proof of Lemma 3.1 since Φ is continuous on $\overline{\mathbb{C}^-}$ (namely via Lemma 7.7), we have for all $z \in \mathbb{C}^+$

$$\begin{aligned}\mathbb{E}[\rho_z^N(\lambda)] &= \mathbb{E}[\Phi(\lambda G(z)_{11})] = \mathbb{E}\left[\Phi\left(\frac{\lambda}{z - a_{11} - \mathbf{a}_1^* G_1(z) \mathbf{a}_1}\right)\right] \\ &= \mathbb{E}\left[\Phi\left(\frac{\lambda}{z - \sum_{j=1}^{N-2} \mathbf{a}_1(j)^2 G_1(z)_{jj}}\right)\right] + o(1).\end{aligned}$$

Then we use Hypothesis made at (22) to get for $\Im z > 0$

$$\begin{aligned}\mathbb{E}[\rho_z^N(\lambda)] &= \int_0^\infty g(y) \mathbb{E}[e^{i\frac{y}{\lambda}(z - \sum_j \mathbf{a}_1(j)^2 G_1(z)_{jj})}] dy + o(1) \\ &= \int_0^\infty g(y) e^{i\frac{y}{\lambda}z} \mathbb{E}\left[\prod_{j=1}^{N-1} \phi_N\left(\frac{y}{\lambda} G_1(z)_{jj}\right)\right] dy + o(1).\end{aligned}$$

Using the definition of Φ and the fact that we assumed that it is bounded on every compact subset (since ρ^N has non positive real part we can cut the integral to keep y bounded up to a small error, as in the previous sections), we have

$$\begin{aligned}\mathbb{E}[\rho_z^N(\lambda)] &= \int_0^\infty g(y) e^{i\frac{y}{\lambda}z} \mathbb{E}\left[e^{\frac{1}{N} \sum_j \Phi(\frac{y}{\lambda} G_1(z)_{jj})}\right] dy + o(1) \\ &= \lambda \int_0^\infty g(\lambda y) e^{iyz} \mathbb{E}\left[e^{\rho_z^{N-1}(y)}\right] dy + o(1).\end{aligned}$$

Now, notice that by Lemmas 7.2 and 7.3, ρ_z^{N-1} can be replaced by $\mathbb{E}[\rho_z^N]$ in the above formula. Moreover, as Φ is uniformly continuous, so is $\mathbb{E}[\rho_z^N]$, so that we can take limit points (or we can use Corollary 3.3 up to assuming (14)) and check that they satisfy (25).

Let us now prove that there is a unique solution to this equation which is analytic in \mathbb{C}^+ and with non positive real part. Suppose that there are two such solutions $\rho_z(\lambda), \tilde{\rho}_z(\lambda)$. For z fixed, let us define $\Delta(\lambda) := |\rho_z(\lambda) - \tilde{\rho}_z(\lambda)|$. Then for all λ , we have, by the hypothesis made on g in (23),

$$\begin{aligned}\Delta(\lambda) &\leq \lambda \int_0^\infty |g(\lambda y)| e^{-y\Im z} \Delta(y) dy \\ &\leq K \left(\lambda^{\gamma+1} \int_0^\infty y^\gamma e^{-y\Im z} \Delta(y) dy + \lambda^{\kappa+1} \int_0^\infty y^\kappa e^{-y\Im z} \Delta(y) dy \right).\end{aligned}$$

It follows that $I_1 := \int_0^\infty \lambda^\gamma e^{-\lambda\Im z} \Delta(\lambda) d\lambda$ and $I_2 := \int_0^\infty \lambda^\kappa e^{-\lambda\Im z} \Delta(\lambda) d\lambda$ satisfy

$$\begin{aligned}I_1 &\leq K \left(I_1 \int_0^\infty \lambda^{2\gamma+1} e^{-\lambda\Im z} d\lambda + I_2 \int_0^\infty \lambda^{\gamma+\kappa+1} e^{-\lambda\Im z} d\lambda \right), \\ I_2 &\leq K \left(I_1 \int_0^\infty \lambda^{\gamma+\kappa+1} e^{-\lambda\Im z} d\lambda + I_2 \int_0^\infty \lambda^{2\kappa+1} e^{-\lambda\Im z} d\lambda \right).\end{aligned}$$

For $\Im z$ large enough, the integrals above are strictly less than $\frac{1}{2K}$, so that $I_1 = I_2 = 0$. It follows that for any fixed λ , $\rho_z(\lambda)$ and $\tilde{\rho}_z(\lambda)$ are analytic functions of z which coincide for $\Im z$ large enough, hence they are equal.

5.2. Fixed point characterization of $\rho_{z,z'}(\cdot, \cdot)$: proof of Theorem 1.13. We now find a fixed point system of equations for the non random function of Corollary 3.3. For $\lambda/\Im z + \lambda'/\Im z' \geq 0$, we set

$$(87) \quad \rho_{z,z'}^{N,k,1}(\lambda, \lambda') = \frac{1}{k-1} \sum_{j=1}^{k-1} \Phi(\lambda G_k(z)_{jj} + \lambda' G'_k(z')_{jj})$$

$$(88) \quad \rho_{z,z'}^{N,k,2}(\lambda, \lambda') = \frac{1}{N-k-1} \sum_{j=k}^{N-1} \Phi(\lambda G_k(z)_{jj} + \lambda' G'_k(z')_{jj}),$$

where we recall that G_k and G'_k are as in (19). To simplify the notations below, as in the previous section, we denote (G, G') instead of (G_k, G'_k) , even though their distribution depends on k .

In the sequel we fix, as in Section 3.3, a number $u \in (0, 1)$ and will give limits in the regime where $N \rightarrow \infty$, $k \rightarrow \infty$ and $k/N \rightarrow u$. We shall then prove that, under the hypotheses of Theorem 1.13 that we assume throughout this section, $(\rho_{z,z'}^{N,k,1}, \rho_{z,z'}^{N,k,2})$ converges almost surely and that its limit satisfies a fixed point system of equations which has a unique analytic solution with non positive real part. The convergence could be shown with minor modifications of Lemma 3.2 under assumption (14), but we do not need this since we work with stronger assumptions. Using the concentration lemma 7.4 (note that Φ is not Lipschitz but can be approximated by Lipschitz functions uniformly on compacts), it is sufficient to prove the fixed point equation for the expectation of these parameters. Moreover, by exchangeability of the k first entries and $N-k$ last entries

$$\begin{aligned} \mathbb{E}[\rho_{z,z'}^{N,k,1}(t, s)] &= \mathbb{E}[\Phi(tG(z)_{11} + sG'(z')_{11})] + o(1), \\ \mathbb{E}[\rho_{z,z'}^{N,k,2}(t, s)] &= \mathbb{E}[\Phi(tG(z)_{NN} + sG'(z')_{NN})] + o(1). \end{aligned}$$

These functions are analytic in $\Lambda = \{z, z' : t/\Im z + s/\Im z' > 0\}$ and uniformly bounded continuous in (t, s) (by uniform continuity of Φ and boundedness of $G(z)$), and hence tight by Arzela-Ascoli on compacts of Λ . We let $\rho_{z,z'}^{u,i}, i = 1, 2$ be a limit point. Since $\mathbb{E}[\rho_{z,z'}^{N,k,i}(t, s)]$ is uniformly bounded on compacts of Λ , the limit points $\rho^{u,i}, i = 1, 2$ are analytic by Montel's theorem. We assume for simplicity that both z and z' have positive imaginary parts (in the general case, one only has to replace $\int_{s=0}^{+\infty} \int_{t=0}^{+\infty}$ below by $\int_{s=0}^{\text{sgn}(\Im z)\infty} \int_{t=0}^{\text{sgn}(\Im z')\infty}$). Under Hypothesis made at (28), we can write by Schur complement formula and getting rid of the off diagonal terms (note that all integrals are finite as they

contain exponentially decreasing terms)

$$\begin{aligned}
\mathbb{E}[\rho_{z,z'}^{N,k,1}(t,s)] &= \int_0^\infty \int_0^\infty e^{i\frac{v}{t}z + i\frac{v'}{s}z'} \mathbb{E}[e^{-i\frac{v}{t}\sum_\ell \mathbf{a}_1(\ell)^2 G_1(z)_{\ell\ell} - i\frac{v'}{s}\sum_\ell \mathbf{a}'_1(\ell)^2 G'_1(z)_{\ell\ell}}] d\tau(v, v') \\
&\quad + \int e^{i\frac{v}{t}} \mathbb{E}[e^{-i\frac{v}{t}\sum_\ell \mathbf{a}_1(\ell)^2 G_1(z)_{\ell\ell}}] d\mu(v) + \int e^{i\frac{v}{t}} \mathbb{E}[e^{-i\frac{v}{s}\sum_\ell \mathbf{a}_1(\ell)^2 G_k(z')_{\ell\ell}}] d\mu(v) + o(1) \\
&= \int_0^\infty \int_0^\infty e^{i\frac{v}{t}z + i\frac{v'}{s}z'} e^{u\mathbb{E}\rho_{z,z'}^{N,k,1}(\frac{v}{t}, \frac{v'}{s}) + (1-u)\rho_z(\frac{v}{t}) + (1-u)\rho_{z'}(\frac{v'}{s})} d\tau(v, v') \\
&\quad + \int_0^\infty e^{i\frac{v}{t}} e^{\rho_z(\frac{v}{t})} d\mu(v) + \int_0^\infty e^{i\frac{v}{s}} e^{\rho_{z'}(\frac{v}{s})} d\mu(v) + o(1)
\end{aligned}$$

and

$$\begin{aligned}
\mathbb{E}[\rho_{z,z'}^{N,k,2}(t,s)] &= \int_0^\infty \int_0^\infty e^{i\frac{v}{t}z + i\frac{v'}{s}z'} \mathbb{E}[e^{-i\frac{v}{t}\sum_\ell \mathbf{a}_N(\ell)^2 G_N(z)_{\ell\ell} - i\frac{v'}{s}\sum_\ell \mathbf{a}'_N(\ell)^2 G'_N(z)_{\ell\ell}}] d\tau(v, v') \\
&\quad + \int e^{i\frac{v}{t}} \mathbb{E}[e^{-i\frac{v}{t}\sum_\ell \mathbf{a}_N(\ell)^2 G_N(z)_{\ell\ell}}] d\mu(v) + \int e^{i\frac{v}{t}} \mathbb{E}[e^{-i\frac{v}{s}\sum_\ell \mathbf{a}_N(\ell)^2 G_k(z')_{\ell\ell}}] d\mu(v) + o(1) \\
&= \int_0^\infty \int_0^\infty e^{i\frac{v}{t}z + i\frac{v'}{s}z'} e^{\rho_z(\frac{v}{t}) + \rho_{z'}(\frac{v'}{s})} d\tau(v, v') \\
&\quad + \int_0^\infty e^{i\frac{v}{t}} e^{\rho_z(\frac{v}{t})} d\mu(v) + \int_0^\infty e^{i\frac{v}{s}} e^{\rho_{z'}(\frac{v}{s})} d\mu(v) + o(1)
\end{aligned}$$

where we used that $\mathbf{a}_1(\ell) = \mathbf{a}'_1(\ell)$ for all $\ell \leq k$ and are independent for $\ell > k$ whereas $\mathbf{a}_N(\ell) = \mathbf{a}'_N(\ell)$ are independent for all ℓ , that ρ_z^N converges towards ρ_z , and that

$$\begin{aligned}
\frac{1}{k-2} \sum_{j=1}^{k-2} \Phi(\lambda G_1(z)_{jj} + \lambda' G'_1(z')_{jj}) &\sim \rho_{z,z'}^{N,k,1} \\
\frac{1}{N-k-2} \sum_{j=k}^{N-2} \Phi(\lambda G_N(z)_{jj} + \lambda' G'_N(z')_{jj}) &\sim \rho_{z,z'}^{N,k,2},
\end{aligned}$$

by Lemma 7.2 (by continuity of Φ , it can be approximated by Lipschitz functions). Hence we find that the limit points $\rho_{z,z'}^{u,1}, \rho_{z,z'}^{u,2}$ of $\rho_{z,z'}^{N,k,1}, \rho_{z,z'}^{N,k,2}$ satisfy (34) and (35). Moreover,

$$\rho_{z,z'}^{N,k} = u\rho_{z,z'}^{N,k,1} + (1-u)\rho_{z,z'}^{N,k,2} + o(1)$$

gives

$$\rho_{z,z'}^u = u\rho_{z,z'}^{u,1} + (1-u)\rho_{z,z'}^{u,2}.$$

Uniqueness under assumption (28): Let $\rho_{z,z'}^{u,1}(t,s)$ and $\tilde{\rho}_{z,z'}^{u,1}(t,s)$ be solutions of Equation (34) with non positive real parts (note here that $\rho_z(\cdot)$ is given and μ is so that the above integrals are finite; hence the last two terms in both equations play the role of a

finite given function).

$$\begin{aligned}
\Delta(t, s) &:= |\rho_{z, z'}^{u, 1}(t, s) - \tilde{\rho}_{z, z'}^{u, 1}(t, s)| \\
&\leq 2 \int_0^\infty \int_0^\infty e^{-\Im z v t^{-1} - \Im z' v' s^{-1}} \frac{d|\tau|(v, v')}{dv dv'} \Delta\left(\frac{v}{t}, \frac{v'}{s}\right) dv dv' \\
&\leq 2ts \int_0^\infty \int_0^\infty e^{-\Im z v - \Im z' v'} K((vt)^\gamma \mathbb{1}_{vt \in [0, 1]} + (vt)^\kappa \mathbb{1}_{vt \in [1, \infty[}) \\
&\quad \times ((v's)^\gamma \mathbb{1}_{v's \in [0, 1]} + (v's)^\kappa \mathbb{1}_{v's \in [1, \infty[}) \Delta(v, v') dv dv' \\
&\leq 2Kts((ts)^\gamma I_\Delta(\gamma, \gamma') + t^\gamma s^\kappa I_\Delta(\gamma, \kappa) + (ts)^\gamma I_\Delta(\kappa, \gamma) + (ts)^\kappa I_\Delta(\kappa, \kappa)),
\end{aligned}$$

where for $(\alpha, \alpha') \in \{\gamma, \kappa\}^2$, we have set

$$I_\Delta(\alpha, \alpha') := \int_0^\infty \int_0^\infty e^{-\Im z v - \Im z' v'} \Delta(v, v') v^\alpha v'^{\alpha'} dv dv'.$$

We put $I(\alpha, \alpha') = I_1(\alpha, \alpha')$ where 1 denote the constant function equal to one. We get after integrating both sides

$$\begin{aligned}
I_\Delta(\alpha, \alpha') &\leq K(I(\alpha + \gamma + 1, \alpha' + \gamma' + 1)I_\Delta(\gamma, \gamma) + I(\alpha + \gamma + 1, \alpha' + \kappa' + 1)I_\Delta(\gamma, \kappa) + \\
&\quad I(\alpha + \kappa + 1, \alpha' + \gamma' + 1)I_\Delta(\kappa, \gamma) + I(\alpha + \kappa + 1, \alpha' + \kappa + 1)I_\Delta(\kappa, \kappa)).
\end{aligned}$$

We consider $\Im z, \Im z'$ large enough so that $I(\alpha + \beta + 1, \alpha' + \beta' + 1) < \frac{1}{4K}$ for any $(\alpha, \alpha'), (\beta, \beta') \in \{\gamma, \kappa\}^2$ to conclude that $\Delta(t, s)$ vanishes then and therefore that $\rho_{z, z'}^{u, s} = \tilde{\rho}_{z, z'}^{u, s}$ for $\Im z$ and $\Im z'$ big enough, $s = 1$ or 2 . By analyticity, we conclude that the system of equations (34) has a unique analytic solution with non positive real part. The functions $\rho^{N, k, 1}$ and $\rho^{N, k, 2}$ are tight and their limit points are characterized by fixed point equations, so they actually converge.

Uniqueness under assumption (27): The limit points of $(\rho_{z, z'}^{N, k, 1}, \rho_{z, z'}^{N, k, 2})$ satisfy (34) and (35) with τ given by (31).

To simplify the notations we assume hereafter $\Im z, \Im z'$ non negative. Notice first that if (g_1, g_2) is a limit point of $(\rho_{z, z'}^{N, k, 1}, \rho_{z, z'}^{N, k, 2})$, g_2 is given and the g_i 's are functions from $(\mathbb{R}^+)^2$ into $\mathcal{L}_{\alpha/2} := \{-re^{i\theta}; r \geq 0, |\theta| \leq \alpha/2\}$, $i = 1, 2$, which are homogeneous of degree $\frac{\alpha}{2}$, i.e. for any $t, s > 0$

$$(89) \quad g_i(t, s) = (t^2 + s^2)^{\frac{\alpha}{4}} g_i\left(\frac{t}{\sqrt{t^2 + s^2}}, \frac{s}{\sqrt{s^2 + t^2}}\right).$$

We show that for any β in $(\frac{\alpha}{2}, 1)$, the system (34) has a unique solution on the set of pair of homogeneous maps $(\mathbb{R}^+)^2 \rightarrow \mathcal{L}_{\alpha/2}$ of degree $\frac{\alpha}{2}$ that satisfy the β -Hölder properties, i.e. have finite $\|\cdot\|_\beta$ norm given by

$$(90) \quad \|g\|_\beta = \max_{(u, v) \in S_+^1} |g(u, v)| + \max_{(u, v) \neq (u', v') \in S_+^1} \frac{|g(u, v) - g(u', v')|}{|(u - u')^2 + (v - v')^2|^{\beta/2}},$$

with $S_+^1 = \{s, t \geq 0, s^2 + t^2 = 1\}$.

Proving that the limit points $(\rho_{z,z'}^{u,1}, \rho_{z,z'}^{u,2})$ of $(\rho_{z,z'}^{N,k,1}, \rho_{z,z'}^{N,k,2})$ are β -Hölder maps for a $\beta > \frac{\alpha}{2}$ allows to conclude the proof of Theorem 1.13. This is the content of the following lemma.

Lemma 5.1. *For any $z, z' \in \mathbb{C}^+$, $u \in [0, 1]$, $i \in \{1, 2\}$ and $\beta \in (\alpha/2, (3\alpha/4) \wedge 1)$,*

$$\|\rho_{z,z'}^{u,i}\|_\beta < \infty.$$

Proof. First, since $|G(z)_{\ell\ell}| \leq \Im z^{-1}$, we have

$$(91) \quad \max_{i=1,2} \max_{(s,t) \in S_+^1} \left| \mathbb{E}[\rho_{z,z'}^{N,k,i}(s,t)] \right| \leq \left(\frac{1}{\Im z} + \frac{1}{\Im z'} \right)^{\frac{\alpha}{2}}.$$

We next show that for any matrix model so that $\Phi(x) = -\sigma(ix)^{\frac{\alpha}{2}}$, for any $2\kappa \in (0, \alpha/2)$

$$(92) \quad \limsup_{N \geq 1} \mathbb{E} \left[\left(\sum |a_{1i}|^2 \right)^{2\kappa} \right] < \infty.$$

Indeed, we can write

$$\mathbb{E} \left[\left(\sum |a_{1i}|^2 \right)^{2\kappa} \right] = c \mathbb{E} \left[\int_0^\infty \frac{1 - e^{-y \sum |a_{1i}|^2}}{y^{1+2\kappa}} dy \right] = \int_0^\infty \frac{1 - \phi_N(-iy)^N}{y^{1+2\kappa}} dy$$

where we have used Fubini for non negative functions. But the above integral is well converging at infinity and we know that ϕ_N converges uniformly on $[0, M]$ for all M finite; hence there exists a finite constant C so that for N large enough

$$\sup_{y \in [0, M]} |y|^{-\frac{\alpha}{2}} N |\phi_N(-iy) - 1| \leq C$$

which yields

$$\mathbb{E} \left[\left(\sum |a_{1i}|^2 \right)^{2\kappa} \right] \leq C \left(1 + \int_0^M \frac{1 - e^{C|y|^{\frac{\alpha}{2}}}}{y^{1+2\kappa}} dy \right) < \infty \text{ for } 2\kappa < \alpha/2.$$

We next show that this estimate implies the β -Hölder property. Indeed, for any $\beta \in [\frac{\alpha}{2}, 1]$, there exists a constant $c = c(\alpha, \beta)$ such that for any x, y in \mathbb{C}^- ,

$$(93) \quad |x^{\frac{\alpha}{2}} - y^{\frac{\alpha}{2}}| \leq c|x - y|^\beta (|x| \wedge |y|)^{\frac{\alpha}{2} - \beta}.$$

Applying this with $x = tG(z)_{jj} + sG'(z')_{jj}$ and $y = t'G(z)_{jj} + s'G'(z')_{jj}$ gives

$$\left| \mathbb{E}[\rho_{z,z'}^{N,k,i}(t,s) - \rho_{z,z'}^{N,k,i}(t',s')] \right| \leq cK_N \left(\frac{1}{\Im z^2} + \frac{1}{(\Im z')^2} \right)^{\frac{\beta}{2}} (|t - t'|^2 + |s - s'|^2)^{\beta/2}$$

with

$$K_N := \mathbb{E} \left[\left(|tG(z)_{11} + sG'(z')_{11}| \wedge |t'G(z)_{11} + s'G'(z')_{11}| \right)^{-\kappa} \right]$$

where $\kappa = \beta - \frac{\alpha}{2} > 0$. It is enough to prove that K_N is uniformly bounded as $\rho_{z,z'}^{u,i}$ is a limit point of $\mathbb{E}[\rho_{z,z'}^{N,k,i}]$. Note that we may assume that $|s - s'| < 1/6$ and $|t - t'| \leq 1/6$ since otherwise the bound is already obtained by (91). But then this implies that either $t, t' \geq 1/4$ or $s, s' \geq 1/4$. Let us assume $t, t' \geq 1/4$. Then, we have

$$|t'G(z)_{11} + s'G'(z')_{11}| \wedge |tG(z)_{11} + sG'(z')_{11}| \geq t' \wedge t \Im G(z)_{11} \geq \frac{1}{4} |\Im G(z)_{11}|.$$

Using Schur formula, we find that

$$\begin{aligned}\mathbb{E}[|\Im G(z)_{11}|^{-\kappa}] &= \mathbb{E}\left[\left(\frac{(\Im(z - \mathbf{a}_1^* G_1(z) \mathbf{a}_1))^2 + (\Re(z - a_{11} - \mathbf{a}_1^* G_1(z) \mathbf{a}_1))^2}{\Im(z - \mathbf{a}_1^* G_1(z) \mathbf{a}_1)}\right)^\kappa\right] \\ &\leq \frac{2^\kappa}{(\Im z)^{3\kappa}} \mathbb{E}\left[\left(C + \left(\sum |a_{1i}|^2\right)^2\right)^\kappa\right]\end{aligned}$$

so that we deduce that for all $\kappa > 0$

$$\mathbb{E}[|\Im G(z)_{11}|^{-\kappa}] \leq \frac{C + \Im z^\kappa}{\Im z^{3\kappa}} \mathbb{E}\left[\left(\left(\sum |a_{1i}|^2\right)^2 + 1\right)^\kappa\right].$$

Equation (92) completes the proof by taking $2\kappa = 2(\beta - \frac{\alpha}{2}) < \frac{\alpha}{2}$. \square

We now prove the uniqueness of the solution of (34) on the set of functions described above. After some change of variables, the equation is equivalent to the following:

$$\begin{aligned}g_1(t, s) &= C_\alpha \int_0^\infty \int_0^\infty \int_0^\infty w^{\alpha/2-1} (w')^{\alpha/2-1} v^{-\alpha/2-1} e^{iW.Z} \\ &\quad \left(e^{ivT.Z + ug_1(W+vT) + (1-u)\rho_{z,z'}(W+vT)} - e^{ug_1(W) + (1-u)g_2(W)} \right) dw dw' dv\end{aligned}$$

where we have denoted in short $W = (w, w')$, $Z = (z, z')$, $T = (s, t)$, $\rho_{z,z'}(W) = \rho_z(w) + \rho_{z'}(w')$ and $W.Z, T.Z$ stand for the scalar products.

After the change of variables $w = r \cos(\theta)$, $w' = r \sin(\theta)$, $v = rv'$, we can rewrite this system of equations as

$$g_1 = F_{z,z'}^u(ug_1 + (1-u)\rho_{z,z'})$$

with, if $T^\dagger = (t, s)$ when $T = (s, t)$,

$$F_{z,z'}^u(g)(T^\dagger) = \int_{v,\theta} \int_{r=0}^\infty r^{\alpha/2-1} e^{ir e_\theta.Z} \left(e^{irvT.Z + r^{\alpha/2}g(e_\theta+vT)} - e^{r^{\alpha/2}g(e_\theta)} \right) dr d\mu(v, \theta),$$

where we have denoted $e_\theta = (\cos(\theta), \sin(\theta))$ and

$$d\mu(v, \theta) = C_\alpha \mathbb{1}_{\theta \in [0, \frac{\pi}{2}]} d\theta (\sin 2\theta)^{\frac{\alpha}{2}-1} \mathbb{1}_{v \in [0, \infty)} v^{-\frac{\alpha}{2}-1}$$

for a constant C_α .

The desired uniqueness follows from the following lemma.

Lemma 5.2. *Let β in $(\frac{\alpha}{2}, 1)$. For any M and for $\Im z, \Im z'$ large enough, the map $F_{z,z'}^u$ is a contraction mapping on the set $\mathcal{C}_{M,\beta}$ of homogenous maps $g : (\mathbb{R}^+)^2 \rightarrow \mathcal{L}_{\alpha/2}$ of degree $\frac{\alpha}{2}$ with β norm bounded by M .*

Proof. To study the Lipschitz property of $F_{z,z'}^u$ as a function of g in $\mathcal{C}_{M,\beta}$ for the norm β , we first set

$$\begin{aligned} F_{z,z'}^u(g)(T^\dagger, \tilde{T}^\dagger) &= F_{z,z'}^u(g)(T^\dagger) - F_{z,z'}^u(g)(\tilde{T}^\dagger) \\ &= \int d\mu(\theta, v) \int_0^\infty dr r^{\alpha/2-1} e^{ir e_\theta \cdot Z} \left(e^{irvT \cdot Z + r^{\alpha/2}g(e_\theta + vT)} - e^{irv\tilde{T} \cdot Z + r^{\alpha/2}g(e_\theta + v\tilde{T})} \right). \end{aligned}$$

We next bound, for given g_1, g_2 in $\mathcal{C}_{M,\beta}$, T_1 in S_+^1 and with T_2 either in S_+^1 or $T_2 = 0$ (which allows to treat in one time two parts of $\|\cdot\|_\beta$)

$$\Delta_{F^u} = |F_{z,z'}^u(g_1)(T_1, T_2) - F_{z,z'}^u(g_2)(T_1, T_2)|.$$

For that task, we shall use some technical estimates, with a constant c that may change from line to line. Remind first the two following bounds from [14, Lem. 5.7] valid for $\gamma > 0$: there exists a constant $c = c(\alpha, \gamma) > 0$ such that for all $h, k \in \mathbb{C}^+$ and for all $x_1, x_2, y_1, y_2 \in \mathcal{L}_{\frac{\alpha}{2}}$,

$$\begin{aligned} (94) \quad J_{\gamma,h}(x_1, x_2, y_1, y_2) &:= \left| \int_0^\infty r^{\gamma-1} e^{irh} \left(\left(e^{r^{\frac{\alpha}{2}}x_1} - e^{r^{\frac{\alpha}{2}}y_1} \right) - \left(e^{r^{\frac{\alpha}{2}}x_2} - e^{r^{\frac{\alpha}{2}}y_2} \right) \right) dr \right| \\ &\leq c \left(|h|^{-\gamma-\frac{\alpha}{2}} |x_1 - x_2 - y_1 + y_2| + |h|^{-\gamma-\alpha} (|x_1 - x_2| + |y_1 - y_2|)(|x_1 - y_1| + |x_2 - y_2|) \right). \end{aligned}$$

Moreover, for all $h, k \in \mathbb{C}^+$, $x, y \in \mathcal{L}_{\frac{\alpha}{2}}$, for $0 < \kappa \leq 1$, we have

$$\begin{aligned} (95) \quad K_{\gamma,h,k}(x, y) &:= \left| \int_0^\infty r^{\gamma-1} (e^{irh} - e^{irk}) \left(e^{r^{\frac{\alpha}{2}}x} - e^{r^{\frac{\alpha}{2}}y} \right) dr \right| \\ &\leq c(|h| \wedge |k|)^{-\gamma-\frac{\alpha}{2}-\kappa} |h - k|^\kappa |x - y|, \end{aligned}$$

Moreover, notice that for (s, t) in S_+^1 , one has $\max(s, t) \geq 1/\sqrt{2}$ and $\max(\cos \theta, \sin \theta) \geq 1/\sqrt{2}$ and thus

$$\Im T \cdot Z \geq \Im z \wedge \Im z' / \sqrt{2} =: \Delta_{z,z'}, \quad \Im(e_\theta \cdot Z) \geq \Delta_{z,z'},$$

so that $|ie_\theta \cdot Z + ivT \cdot Z| \geq \Delta_{z,z'}(1 + 1_{T \in S_+^1} v)$. At last, with $a_i = e_\theta + vT_i$ for $i = 1, 2$, straightforward uses of the β -norm and the inequalities $|a_i| \leq (1 + v)$, $||a_1| - |a_2|| \leq v|T_1 - T_2|$, $|a_i| \geq \frac{1}{\sqrt{2}}(v \vee 1)$ if $T_i \in S_+^1$, $\left| \frac{a_1}{|a_1|} - \frac{a_2}{|a_2|} \right| \leq (v \vee 1)^{-2} |T_1 - T_2| v(1 + v)$, and (93) gives the estimates

$$\begin{aligned} \left| \sum_{i,j=1}^2 (-1)^{i+j} g_i(a_j) \right| &\leq c \|g_1 - g_2\|_\beta |T_1 - T_2|^\beta f_\beta(v), \\ \left(\sum_{i=1}^2 |g_1(a_i) - g_2(a_i)| \right) \left(\sum_{i=1}^2 |g_i(a_1) - g_i(a_2)| \right) &\leq cM \|g_2 - g_1\|_\beta |T_1 - T_2|^\beta (1 + v)^{\frac{\alpha}{2}} f_\beta(v), \end{aligned}$$

where $f_\beta(v) = v^\beta \left((1+v)^{\frac{\alpha}{2}+\beta} (v \vee 1)^{-2\beta} + 1 \right)$. Using this series of estimates, we find that,

$$\begin{aligned}
\Delta_{F^u} &\leq \int K_{\frac{\alpha}{2}, e_\theta, Z+vT, Z, e_\theta, Z+v\tilde{T}, Z} (g(e_\theta + v\tilde{T}), \tilde{g}(e_\theta + v\tilde{T})) d\mu(v, \theta) \\
&\quad + \int J_{\frac{\alpha}{2}, e_\theta, Z+vT, Z} (g(e_\theta + vT), \tilde{g}(e_\theta + vT), g(e_\theta + v\tilde{T}), \tilde{g}(e_\theta + v\tilde{T})) d\mu(v, \theta) \\
&\leq c \|g - \tilde{g}\|_\beta \left(\Delta_{z, z'}^{-\alpha} \mathbb{1}_{\tilde{T}=0} \int 1_{v \geq 1} d\mu(\theta, v) \right. \\
&\quad + |\tilde{T} - T|^\beta \Delta_{z, z'}^{-\alpha-\beta} (|z| + |z'|)^\beta \int (1 + v \mathbb{1}_{|\tilde{T}|=1})^{-\beta-\alpha} (1 + \mathbb{1}_{\tilde{T}=0, v \leq 1}) v^\beta d\mu(\theta, v) \\
&\quad \left. + |T - \tilde{T}|^\beta (M \Delta_{z, z'}^{-3\alpha/2} + \Delta_{z, z'}^{-\alpha}) \int v^\beta (1+v)^\beta \left((v \vee 1)^{-2\beta} + 1 \right) d\mu(\theta, v) \right).
\end{aligned}$$

While using (95), we chose $\gamma = \alpha/2$ and $\kappa = \beta$ when $\tilde{T} \in S_1$ or $\tilde{T} = 0, v \leq 1$, $\kappa = 0$ when $\tilde{T} = 0$ and $v \geq 1$. As the integrals are finite we obtain the desired bound for $g, \tilde{g} \in \mathcal{C}_{M, \beta}$

$$\|F_{z, z'}^u(g_1) - F_{z, z'}^u(\tilde{g}_1)\|_\beta \leq C(z, z', M) \|g_1 - \tilde{g}_1\|_\beta,$$

with $C(z, z', M) < 1$ if $\Im z \wedge \Im z'$ is large enough. \square

Taking two solutions of (34) and (35) in $\mathcal{C}_{M, \beta}$, we deduce that they are equal when $\Im z \wedge \Im z'$ is large enough, and thus everywhere by analyticity.

6. PROOFS OF LEMMAS 1.3, 1.8 AND 1.12

6.1. Proof of Lemma 1.3. Let us first treat the case of Wigner matrices with exploding moments. First and second parts of Hypothesis 1.1, as well as (4), are satisfied for $c = c_\varepsilon = 0$. Let us then define ν_N to be the law of $a^2 = a_{11}^2$ and m_N to be the measure with density Nx with respect to ν_N , so that for any test function f , we have

$$\int f dm_N = \int Nx f(x) d\nu_N(x) = N\mathbb{E}[a^2 f(a^2)].$$

Then for each $k \geq 0$,

$$\int x^k dm_N(x) = N\mathbb{E}[|a_{11}|^{2(k+1)}] \longrightarrow C_{k+1}.$$

As there is a unique measure m on \mathbb{R}_+ with $(C_{k+1})_{k \geq 0}$ as sequence of moments, this proves that m_N converges weakly to m . Then (7) is a consequence that for any continuous bounded function f on \mathbb{R}_+ , $\int f dm_N \longrightarrow \int f dm$ and the convergence is uniform on uniformly Lipschitz sets of functions (apply this with $f(x) = \frac{e^{-i\lambda x} - 1}{x}$).

Let us now treat the case of Lévy matrices. Set $t_N := N^\mu \in (0, \frac{1}{2(2-\alpha)})$ and define $b := a \mathbb{1}_{|a| \leq t_N}$ and $c := a \mathbb{1}_{|a| > t_N}$. Then (2) is obvious by Hypothesis (8) and the fact that

$a_N = N^{1/\alpha}$ up to a slowly varying factor and (3) follows directly from Lemma 5.8 of [11] (in fact, this lemma gives us the right upper bound for the second moment of b , which of course implies that it is true for its variance). Let us now treat the second part of the hypothesis. Let us fix $\varepsilon > 0$ and define $b_\varepsilon = a \mathbb{1}_{|a| \leq B}$ (for a constant B which will be specified below) and $c_\varepsilon := a - b_\varepsilon$. For L as in (8), we have

$$\mathbb{P}(c_\varepsilon \neq 0) = \frac{L(a_N B)}{(a_N B)^\alpha} \sim \frac{L(a_N)}{a_N^\alpha B^\alpha} \sim \frac{1}{NB^\alpha}.$$

Hence (5) is satisfied if B is chosen large enough. For the convergence of the truncated even moments, see [38, Sect. 1.2.1]. Moreover, (7) follows from the results of e.g. Section 8.1.3 of [13]. At last, (4) is satisfied for Lévy matrices by e.g. Section 10 of [10].

6.2. Proof of Lemma 1.8. In the case of Lévy matrices, the expression

$$-\sigma(i\lambda)^{\alpha/2} = \int_{y=0}^{+\infty} C_\alpha y^{\frac{\alpha}{2}-1} e^{i\frac{y}{\lambda}} dy \quad (\lambda \in \mathbb{C}^-)$$

relies an application of residues formula which gives, for $z \in \mathbb{C}^+$ and $\alpha > 0$,

$$(96) \quad \Gamma(\alpha/2) = -i \int_{t=0}^{+\infty} (-izt)^{\frac{\alpha}{2}-1} e^{itz} z dt.$$

In the case of Wigner matrices with exploding moment, one first needs to use the following formula, for $\xi \in \mathbb{C}$ with positive real part:

$$(97) \quad 1 - e^{-\xi} = \int_0^{+\infty} \frac{J_1(2\sqrt{t})}{\sqrt{t}} e^{-t/\xi} dt,$$

which is proved in the following way (using (96)):

$$1 - e^{-\xi} = \sum_{p \geq 0} \frac{(-1)^p}{p!(p+1)!} p! \xi^{p+1} = \sum_{p \geq 0} \frac{(-1)^p}{p!(p+1)!} \int_0^{+\infty} t^p e^{-t/\xi} dt = \int_0^{+\infty} \frac{J_1(2\sqrt{t})}{\sqrt{t}} e^{-t/\xi} dt.$$

It follows that for m_N the measure introduced in the proof of Lemma 1.3 above, we have

$$N(\phi_N(\lambda) - 1) = N(\mathbb{E} e^{-i\lambda a^2} - 1) = -N \mathbb{E} \int_0^{+\infty} \frac{J_1(2\sqrt{t})}{\sqrt{t}} e^{-\frac{t}{i\lambda a^2}} dt = \int_0^{+\infty} g_N(y) e^{i\frac{y}{\lambda}} dy$$

with

$$(98) \quad g_N(y) := -N \frac{\mathbb{E}[|a| J_1(2\sqrt{y}|a|)]}{\sqrt{y}} = -N \mathbb{E}[a^2 \frac{J_1(2\sqrt{y}a^2)}{\sqrt{y}a^2}] = \int f_y(x) dm_N(x)$$

for $f_y(x) := -\frac{J_1(2\sqrt{xy})}{\sqrt{xy}}$. As m_N converges weakly to m and f_y is continuous and bounded, we have

$$g_N(y) \longrightarrow - \int \frac{J_1(2\sqrt{xy})}{\sqrt{xy}} dm(x).$$

6.3. Proof of Lemma 1.12. The case of Lévy matrices is obvious. To treat the case of Wigner matrices with exploding moment, first note that by (97), writing

$$e^{-\xi-\xi'} - 1 = (e^{-\xi} - 1)(e^{-\xi'} - 1) + (e^{-\xi} - 1) + (e^{-\xi'} - 1),$$

we have, for any $\xi, \xi' \in \mathbb{C}$ with positive real parts,

$$(99) \quad e^{-\xi-\xi'} - 1 = \iint_{(\mathbb{R}_+)^2} \frac{J_1(2\sqrt{t})J_1(2\sqrt{t'})}{\sqrt{tt'}} e^{-\frac{t}{\xi}-\frac{t'}{\xi'}} dt dt' \\ - \int_0^{+\infty} \frac{J_1(2\sqrt{t})}{\sqrt{t}} e^{-t/\xi} dt - \int_0^{+\infty} \frac{J_1(2\sqrt{t'})}{\sqrt{t'}} e^{-t'/\xi'} dt'$$

As a consequence, for $\lambda, \mu \in \mathbb{C}^-$,

$$\begin{aligned} N(\phi_N(\lambda + \mu) - 1) &= N\mathbb{E}[e^{-i\lambda a^2 - i\mu a^2} - 1] \\ &= \iint_{(\mathbb{R}_+)^2} g_N(u, u') e^{i\frac{u}{\lambda} + i\frac{u'}{\mu}} du du' \\ &\quad + \int_0^{+\infty} g_N(u) e^{iu/\lambda} du + \int_0^{+\infty} g_N(u') e^{iu'/\mu} du' \end{aligned}$$

with $g_N(u)$ defined by (98) and

$$g_N(u, u') := \int \frac{J_1(2\sqrt{ux})J_1(2\sqrt{u'x})}{\sqrt{uu'}} dm_N(x).$$

Then one concludes as for the proof of Lemma 1.8.

7. APPENDIX

7.1. Concentration of random matrices with independent rows and linear algebra lemmas. This section is mostly a reminder of results from [14] and [15].

The total variation norm of $f : \mathbb{R} \rightarrow \mathbb{C}$ is

$$(100) \quad \|f\|_{\text{TV}} := \sup \sum_{k \in \mathbb{Z}} |f(x_{k+1}) - f(x_k)|,$$

where the supremum runs over all sequences $(x_k)_{k \in \mathbb{Z}}$ such that $x_{k+1} \geq x_k$ for any $k \in \mathbb{Z}$. If $f = \mathbb{1}_{(-\infty, s]}$ for some real s then $\|f\|_{\text{TV}} = 1$, while if f is absolutely continuous (hence almost everywhere differentiable and equal to the integral of its derivative) with derivative in $L^1(\mathbb{R})$, we get

$$(101) \quad \|f\|_{\text{TV}} = \int |f'(t)| dt.$$

The next lemma is an easy consequence of Cauchy-Weyl interlacing Theorem. It is an ingredient of the proof of Lemma 7.3.

Lemma 7.1 (Interlacing of eigenvalues). *Let A be an $N \times N$ hermitian matrix and B a principal minor of A . Then for any $f : \mathbb{R} \rightarrow \mathbb{C}$ such that $\|f\|_{TV} \leq 1$ and $\lim_{|x| \rightarrow \infty} f(x) = 0$,*

$$\left| \sum_{i=1}^N f(\lambda_i(A)) - \sum_{i=1}^{N-1} f(\lambda_i(B)) \right| \leq 1.$$

Lemma 7.2. *Let A_1, A_2 be $N \times N$ random Hermitian matrices and \tilde{A}_1, \tilde{A}_2 be $n-1 \times n-1$ matrices obtained from A_1 and A_2 respectively by removing the ℓ -th row and column, for some $\ell \in \{1, \dots, N\}$. Let $z, z' \in \mathbb{C}$, $t, t' \in \mathbb{R}$ so that $\Im z t > 0$ and $\Im z' t' > 0$ and set $G = (z - A_1)^{-1}$, $G' = (z' - A_2)^{-1}$ and $\tilde{G} = (z - \tilde{A}_1)^{-1}$, $\tilde{G}' = (z' - \tilde{A}_2)^{-1}$. Then, for any function f on $B_{z, z', t, t'} := \{g \in \mathbb{C}^-; |g| \leq C(z, z', t, t')\}$ with $C(z, z', t, t') = t(\Im z)^{-1} + t'(\Im z')^{-1}$, we have*

(102)

$$\left| \frac{1}{N} \sum_{k=1}^N f(tG_{kk} + t'G'_{kk}) - \frac{1}{N} \sum_{k=1}^{N-1} f(t\tilde{G}_{kk} + t'\tilde{G}'_{kk}) \right| \leq \frac{C(z, z', t, t')}{N} \|f\|_{\text{Lip}} + \frac{\|f\|_{\infty}}{N},$$

where $\|f\|_{\text{Lip}} := \sup_{x \neq y} \frac{|f(y) - f(x)|}{|y - x|}$ and $\|f\|_{\infty} := \sup_x |f(x)|$, both supremums running over the elements of $B_{z, z', t, t'}$.

Proof. The proof is similar to [14, (91)]. We denote by \bar{A}_1, \bar{A}_2 the $N \times N$ matrices whose entries are the same as A_1, A_2 except for the ℓ th rows and column which have zero entries. We denote \bar{G}, \bar{G}' the corresponding Stieltjes transform. Then, \tilde{G}, \tilde{G}' equal \bar{G}, \bar{G}' except at the ℓ th column and row (where it is equal to $z^{-1}1_{i=j=k}$). Therefore, noting $\bar{M} = t\bar{G}(z) + t'\bar{G}'(z')$ and $\tilde{M} = t\tilde{G}(z) + t'\tilde{G}'(z)$, we conclude that

$$\left| \frac{1}{N} \sum_{k=1}^N f(\tilde{M}_{kk}) - \frac{1}{N} \sum_{k=1}^{N-1} f(\bar{M}_{kk}) \right| \leq \frac{\|f\|_{\infty}}{N}.$$

Moreover, let $M = tG(z) + t'G'(z')$ and note that $A_1 - \bar{A}_1$ and $A_2 - \bar{A}_2$ have rank one so that $M - \bar{M}$ has rank one. On the other hand it is bounded uniformly by $C = C(z, z', t, t')$. Hence, we can write $M - \bar{M} = cuu^*$ with a unit vector u and c bounded by C . Therefore, since f is Lipschitz,

$$\begin{aligned} \left| \frac{1}{N} \sum_{k=1}^N f(M_{kk}) - \frac{1}{N} \sum_{k=1}^{N-1} f(\bar{M}_{kk}) \right| &\leq \frac{\|f\|_{\text{Lip}}}{N} \sum_{k=1}^N |M_{kk} - \bar{M}_{kk}| \leq \frac{\|f\|_{\text{Lip}}}{N} \sum_{k=1}^N C \langle e_k, u \rangle^2 \\ &= \frac{C\|f\|_{\text{Lip}}}{N}. \end{aligned}$$

□

Lemma 7.3 (Concentration for spectral measures [16]). *Let A be an $N \times N$ random Hermitian matrix. Let us assume that the vectors $(A_i)_{1 \leq i \leq N}$, where $A_i := (A_{ij})_{1 \leq j \leq i} \in \mathbb{C}^i$, are independent. Then for any measurable $f : \mathbb{R} \rightarrow \mathbb{C}$ such that $\mathbb{E}|\int f d\mu_A| < \infty$, and*

every $t \geq 0$,

$$\mathbb{P} \left(\left| \int f d\mu_A - \mathbb{E} \int f d\mu_A \right| \geq t \right) \leq \exp \left(-\frac{Nt^2}{2\|f\|_{\text{TV}}^2} \right).$$

As a consequence,

$$(103) \quad \mathbb{E} \left[\left| \int f d\mu_A - \mathbb{E} \int f d\mu_A \right|^2 \right] \leq 2 \frac{\|f\|_{\text{TV}}^2}{N}.$$

Lemma 7.4 (Concentration for the diagonal of the resolvent). *a) Let A be an $N \times N$ random Hermitian matrix and consider its resolvent matrix $G(z) = (A - z)^{-1}$, $z \in \mathbb{C}_+$. Let us assume that the vectors $(A_i)_{1 \leq i \leq N}$, where $A_i := (A_{ij})_{1 \leq j \leq i} \in \mathbb{C}^i$, are independent. Then for any $f : \mathbb{C}^- \rightarrow \mathbb{R}$ such that $\|f\|_{\text{Lip}} \leq 1$, and every $t \geq 0$,*

$$\mathbb{P} \left(\left| \frac{1}{n} \sum_{k=1}^N f(G(z)_{kk}) - \mathbb{E} \frac{1}{N} \sum_{k=1}^N f(G(z)_{kk}) \right| \geq t \right) \leq 2 \exp \left(-\frac{N\Im(z)^2 t^2}{8} \right).$$

b) Let A' be an $N \times N$ self-adjoint matrices so that $A'_{ij} = A_{ij}$, $i \wedge j \leq k$ and $(A'_{ij})_{j \geq k+1, i \geq k+1}$ is independent from $(A_{ij})_{j \geq k+1, i \geq k+1}$ but with the same distribution. Let $G(z) = (z - A)^{-1}$ and $G'(z) = (z - A')^{-1}$ and set for a Lipschitz function f on $\overline{\mathbb{C}}^-$,

$$\begin{aligned} \rho_{z,z'}^{N,k,1}(\lambda, \lambda')[f] &:= \frac{1}{k} \sum_{\ell=1}^k f(\lambda G(z)_{\ell\ell} + \lambda' G(z')_{\ell\ell}) \\ \rho_{z,z'}^{N,k,2}(\lambda, \lambda')[f] &:= \frac{1}{N-k} \sum_{\ell=k+1}^N f(\lambda G(z)_{\ell\ell} + \lambda' G(z')_{\ell\ell}) \end{aligned}$$

Then, for $\lambda/\Im z \geq 0$, $\lambda'/\Im z' \geq 0$, we have for all $\delta \geq 0$, $s \in \{0, 1\}$,

$$(104) \quad \mathbb{P} \left(\left| \rho_{z,z'}^{N,k,s+1}(\lambda, \lambda')[f] - \mathbb{E}[\rho_{z,z'}^{N,k,s+1}(\lambda, \lambda')[f]] \right| \geq \delta \right) \leq 2e^{-\frac{\delta^2((k-1)^{1-s} + (N-k-1)^s)}{8\|f\|_{\text{Lip}}^2 C(\lambda, \lambda', t, t')^2}}$$

with

$$C(\lambda, \lambda', z, z') = \frac{2\lambda}{\Im z} + \frac{2t\lambda'}{\Im z'}.$$

Proof. The first point is proved as in [14, Lemma C.3]. We outline the proof of the second point which is very similar to [14, Lemma C.3]. We concentrate on $\rho_{\lambda, \lambda'}^{N,k,1}$, the other case being similar. By Azuma-Hoeffding inequality, it is sufficient to show that

$$X_p := \mathbb{E}[\rho_{z,z'}^{N,k,1}(\lambda, \lambda')[f] | \mathcal{F}_p] - \mathbb{E}[\rho_{z,z'}^{N,k,1}(\lambda, \lambda')[f] | \mathcal{F}_{p-1}]$$

is uniformly bounded by $\|f\|_{\text{Lip}} C(\lambda, \lambda', t, t') k^{-1}$. Here \mathcal{F}_p is the σ -algebra generated with respect to the p first column (and row) vectors. Note that X_p can be written as the conditional expectation of the difference of the parameter f evaluated at two sets A, A'

and \tilde{A}, \tilde{A}' which differ only at the p -th vector column (and row). Hence, we may follow the proof of Lemma 7.2 to conclude that

$$|X_p| \leq \frac{\|f\|_{\text{Lip}}}{k} \sum_{\ell=1}^k |(M - \tilde{M})_{\ell\ell}| = \frac{|c|\|f\|_{\text{Lip}}}{k} \sum_{\ell=1}^k \langle u, e_\ell \rangle^2 \leq \frac{|c|\|f\|_{\text{Lip}}}{k}.$$

□

Let $H = [h_{ij}]$ be an $N \times N$ Hermitian matrix and $z \in \mathbb{C} \setminus \mathbb{R}$. Define $G := (z - H)^{-1}$.

Lemma 7.5 (Difference of traces of a matrix and its major submatrices). *Let H_k be the submatrix of H obtained by removing its k -th row and its k -th column and set $G_k := (z - H_k)^{-1}$. Let also \mathbf{a}_k be the k -th column of H where the k -th entry has been removed. Then*

$$(105) \quad \text{Tr}(G) - \text{Tr}(G_k) = \frac{1 + \mathbf{a}_k^* G_k^2 \mathbf{a}_k}{z - h_{kk} - \mathbf{a}_k^* G_k \mathbf{a}_k}.$$

Moreover,

$$(106) \quad |\text{Tr}(G) - \text{Tr}(G_k)| \leq \pi |\Im z|^{-1}.$$

Proof. For (105), see [4, Th. A.5]. For (106), see Lemma 7.1. □

Lemma 7.6. *With the notation introduced above the previous lemma, for each $1 \leq j \leq N$,*

$$(107) \quad \Im z \times \Im G_{jj} < 0,$$

$$(108) \quad |\Im z| \times |(G^2)_{jj}| \leq |\Im G_{jj}| \leq |\Im z|^{-1}$$

and for any $\mathbf{a} = (\mathbf{a}_1, \dots, \mathbf{a}_N) \in \mathbb{C}^N$,

$$(109) \quad \left| \frac{1 + \sum_j |\mathbf{a}_j|^2 (G^2)_{jj}}{z - \sum_j |\mathbf{a}_j|^2 G_{jj}} \right| \leq 2 |\Im z|^{-1}.$$

Proof. Set $z = x + iy$, $x, y \in \mathbb{R}$. Let $\lambda_1, \dots, \lambda_N$ be the eigenvalues of H , associated with the orthonormalized collection of eigenvectors $\mathbf{u}_1, \dots, \mathbf{u}_N$. Let also \mathbf{e}_j denote the j th vector of the canonical basis. Then (107) and (108) follow directly from the following:

$$(110) \quad |(G^2)_{jj}| = \left| \sum_{k=1}^N \frac{|\langle \mathbf{e}_j, \mathbf{u}_k \rangle|^2}{(z - \lambda_k)^2} \right| \leq \sum_{k=1}^N \frac{|\langle \mathbf{e}_j, \mathbf{u}_k \rangle|^2}{(x - \lambda_k)^2 + y^2} \quad \Im G_{jj} = - \sum_{k=1}^N \frac{|\langle \mathbf{e}_j, \mathbf{u}_k \rangle|^2 y}{(x - \lambda_k)^2 + y^2}$$

Let us now prove (109). By (107), we know that $\Im z$ and $-\Im G_{jj}$ have the same sign, so

$$(111) \quad \left| \frac{1}{z - \sum_j |\mathbf{a}_j|^2 G_{jj}} \right| \leq \frac{1}{|\Im(z - \sum_j |\mathbf{a}_j|^2 G_{jj})|} \leq |\Im z|^{-1}.$$

Hence it remains only to prove (109). This is a direct consequence of (107) and (108) which imply the second and last inequality

$$\left| \frac{\sum_j |\mathbf{a}_j|^2 (G^2)_{jj}}{z - \sum_j |\mathbf{a}_j|^2 G_{jj}} \right| \leq \frac{\sum_j |\mathbf{a}_j|^2 |(G^2)_{jj}|}{|\Im(z - \sum_j |\mathbf{a}_j|^2 G_{jj}(z))|} \leq \frac{\sum_j |\mathbf{a}_j|^2 |(G^2)_{jj}|}{\sum_j |\mathbf{a}_j|^2 |\Im G_{jj}|} \leq \frac{1}{\Im z}.$$

□

7.2. Vanishing of non diagonal terms in certain quadratic sums of random vectors. Let $\|M\|_{\text{op}}$ denote the operator norm of a complex matrix with respect to the canonical Hermitian norms.

Lemma 7.7. *For each $N \geq 1$, let (a_1, \dots, a_N) be a family of i.i.d. copies of a random variable a such that a can be decomposed into $a = b + c$ with b, c such that b is centered and (2), (3) of Hypothesis 1.1 are satisfied. Let also B_N be a non random $N \times N$ matrix such that $N^{-1} \text{Tr}(B_N B_N^*)$ is bounded. Then we have the convergence in probability*

$$X := \sum_{i \neq j} a_i B_{ij} a_j \longrightarrow 0.$$

Proof. For each i , let $a_i = b_i + c_i$ be the decomposition of a_i corresponding to $a = b + c$. Set $X^b := \sum_{i \neq j} b_i B_{ij} b_j$ and define the event $E_N := \{\forall i, c_i = 0\}$. Note that when E_N holds, $X = X^b$. But by (2) and the union bound, $\mathbb{P}(E_N) \longrightarrow 1$, so that it suffices to prove that X^b converges in probability to zero, which follows from the fact that its second moment tends to zero. Indeed, by independence of the b_i 's and the fact that they are centered, its second moment is

$$\mathbb{E} \sum_{i \neq j} b_i^2 (B_{ij}^2 + B_{ij} B_{ji}) b_j^2 \leq 2N \text{Var}(b)^2 \frac{1}{N} \text{Tr}(B_N B_N^*).$$

□

7.3. CLT for martingales. Let $(\mathcal{F}_k)_{k \geq 0}$ be a filtration such that $\mathcal{F}_0 = \{\emptyset, \Omega\}$ and let $(M_k)_{k \geq 0}$ be a square-integrable complex-valued martingale starting at zero with respect to this filtration. For $k \geq 1$, we define the random variables

$$Y_k := M_k - M_{k-1} \quad v_k := \mathbb{E}[|Y_k|^2 | \mathcal{F}_{k-1}] \quad \tau_k := \mathbb{E}[Y_k^2 | \mathcal{F}_{k-1}]$$

and we also define

$$v := \sum_{k \geq 1} v_k \quad \tau := \sum_{k \geq 1} \tau_k \quad L(\varepsilon) := \sum_{k \geq 1} \mathbb{E}[|Y_k|^2 \mathbb{1}_{|Y_k| \geq \varepsilon}].$$

Let now everything depend on a parameter N , so that $\mathcal{F}_k = \mathcal{F}_k(N)$, $Y_k = Y_k(N)$, $v = v(N)$, $\tau = \tau(N)$, $L(\varepsilon) = L(\varepsilon, N), \dots$

Then we have the following theorem. It is proved in the real case at [12, Th. 35.12]. The complex case can be deduced noticing that for $z \in \mathbb{C}$, $\Re(z)^2$, $\Im(z)^2$ and $\Re(z)\Im(z)$ are linear combinations of z^2 , \bar{z}^2 , $|z|^2$.

Theorem 7.8. *Suppose that for some constants $v \geq 0, \tau \in \mathbb{C}$, we have the convergence in probability for any $\varepsilon > 0$*

$$v(N) \xrightarrow[N \rightarrow \infty]{} v \quad \tau(N) \xrightarrow[N \rightarrow \infty]{} \tau, \quad L(\varepsilon, N) \xrightarrow[N \rightarrow \infty]{} 0.$$

Then we have the convergence in distribution

$$\sum_{k \geq 1} Y_k(N) \xrightarrow[N \rightarrow \infty]{} Z,$$

where Z is a centered complex Gaussian variable such that $\mathbb{E}(|Z|^2) = v$ and $\mathbb{E}(Z^2) = \tau$.

7.4. Extension of CLTs for random matrices. The following lemma is borrowed from the paper of Shcherbina and Tirozzi [43], except that we do not require, in the hypotheses here, V to be continuous, which is very useful in our case.

Lemma 7.9. *Let, for each N , $(\xi_i^{(N)})_{i=1}^N$ be a collection of \mathbb{R}^d -valued random variables. For each $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}$, set*

$$Z_N(\varphi) := u_N \sum_{i=1}^N (\varphi(\xi_i^{(N)}) - \mathbb{E}[\varphi(\xi_i^{(N)})]),$$

where u_N is a sequence of real numbers. We make the following hypotheses :

- For any φ in a certain normed subspace $(\mathcal{L}, \|\cdot\|)$ of the set of functions $\mathbb{R}^d \rightarrow \mathbb{R}$,

$$(112) \quad \mathbb{E}[Z_N(\varphi)^2] \leq \|\varphi\|^2.$$

- There is a dense subspace $\mathcal{L}_1 \subset \mathcal{L}$ and a quadratic form $V : \mathcal{L}_1 \rightarrow \mathbb{R}_+$ such that for any $\varphi \in \mathcal{L}_1$, we have the convergence in distribution

$$(113) \quad Z_N(\varphi) \xrightarrow[N \rightarrow \infty]{} \mathcal{N}(0, V(\varphi)).$$

Then V is continuous on \mathcal{L}_1 , can be (uniquely) continuously extended to \mathcal{L} and (113) is true for any $\varphi \in \mathcal{L}$.

Proof. This is exactly Proposition 4 of [43], except that in [43], the hypotheses include the continuity of V . Let us prove that the hypotheses made here imply that V is continuous on \mathcal{L}_1 . For any $\varphi \in \mathcal{L}_1$, $V(\varphi)$ is the second moment of the limit law of $Z_N(\varphi)$. Hence

$$V(\varphi) \leq \liminf_{N \rightarrow \infty} \mathbb{E}[Z_N(\varphi)^2] \leq \|\varphi\|^2.$$

This proves that the quadratic form V is continuous. □

7.5. On the Hadamard product of Hermitian matrices.

Proposition 7.10. *Let A_1, \dots, A_p be N by N Hermitian random matrices whose entries have all their moments. Then, there exists a family of random variables (a_1, \dots, a_p) whose joint distribution is given by:*

$$\mathbb{E}[a_1^{n_1} \dots a_p^{n_p}] = \mathbb{E} \left[\frac{1}{N} \text{Tr}[A_1^{n_1} \circ \dots \circ A_p^{n_p}] \right], \quad \forall n_1, \dots, n_p \geq 0,$$

where \circ denotes the Hadamard (entry-wise) product.

Proof. Step 1. We first assume that the matrices are deterministic and have distinct eigenvalues. By the spectral decomposition, for $j = 1, \dots, p$, we have $A_j = \sum_{i=1}^N \lambda_{j,i} u_{j,i} u_{j,i}^*$ where $\Lambda_j = (\lambda_{j,i})_{i=1, \dots, N}$ is the family of eigenvalues of A_j arranged in increasing order, and $U_j = (u_{j,i})_{i=1, \dots, N}$ is the family of associated eigenvectors. For any $n_1, \dots, n_p \geq 0$, one has

$$\begin{aligned} \frac{1}{N} \text{Tr}[A_1^{n_1} \circ \dots \circ A_p^{n_p}] &= \frac{1}{N} \sum_{k=1}^N (A_1^{n_1})(k, k) \dots (A_p^{n_p})(k, k) \\ &= \frac{1}{N} \sum_{k=1}^N \left(\sum_{i_1=1}^N \lambda_{1,i_1}^{n_1} u_{1,i_1} u_{1,i_1}^* \right)(k, k) \dots \left(\sum_{i_p=1}^N \lambda_{p,i_p}^{n_p} u_{p,i_p} u_{p,i_p}^* \right)(k, k) \\ &= \frac{1}{N^p} \sum_{i_1, \dots, i_p=1}^N \lambda_{1,i_1}^{n_1} \dots \lambda_{p,i_p}^{n_p} \times \left(N^{p-1} \sum_{k=1}^N |u_{1,i_1}(k)|^2 \dots |u_{p,i_p}(k)|^2 \right). \end{aligned}$$

For $j = 1, \dots, p$, we set $d\mu_{\Lambda_j} = \frac{1}{N} \sum_{i=1}^N \delta_{\lambda_{j,i}}$ the empirical eigenvalues distributions of A_j . We denote $F_{\Lambda_j}(t) = \mu_{\Lambda_j}((-\infty, t])$ the cumulative function of $d\mu_{\Lambda_j}$. Since the eigenvalues of the matrices are distinct, one has $F_{\Lambda_j}(\lambda_{j,i}) = \frac{i}{N}$ for any $i = 1, \dots, N$, $j = 1, \dots, p$. Hence, we have

$$\frac{1}{N} \text{Tr}[A_1^{n_1} \circ \dots \circ A_p^{n_p}] = \int_{\mathbb{R}^p} \lambda_1^{n_1} \dots \lambda_p^{n_p} f_N((\lambda_j, \Lambda_j, U_j)_{j=1, \dots, p}) d\mu_{\Lambda_1}(\lambda_1) \dots d\mu_{\Lambda_p}(\lambda_p),$$

where $f_N((\lambda_j, \Lambda_j, U_j)_{j=1, \dots, p}) = \left(N^{p-1} \sum_{k=1}^N \prod_{j=1}^p |u_{j, (NF_{\Lambda_j}(\lambda_j))}(k)|^2 \right)$. Hence, a family of random variables (a_1, \dots, a_p) as in the proposition exists and its joint distribution has density $f_N((\cdot, \Lambda_j, U_j)_{j=1, \dots, p})$ with respect to $\mu_{\Lambda_1} \otimes \dots \otimes \mu_{\Lambda_p}$.

Step 2. We now assume that (A_1, \dots, A_p) are random and that their joint distributions have a density with respect to the Lebesgue measure on \mathcal{H}_N^p , where \mathcal{H}_N is the space of Hermitian matrices of size N . In particular, the matrices have almost surely N distinct eigenvalues, see [21]. The spectral decompositions of the previous step are measurable (see [21, Section 5.3]) and, with the above notations (Λ_j, U_j) for eigenvalues and eigenvectors of A_j , we can write the joint distribution of (A_1, \dots, A_p) in the form $g_N((\Lambda_j, U_j)_{j=1, \dots, p}) d\mu_{\Delta_N}(\Lambda_1) \dots d\mu_{\Delta_N}(\Lambda_p) d\mu_{\mathcal{U}_N}(U_1) \dots d\mu_{\mathcal{U}_N}(U_p)$. The symbol μ_{Δ_N} denotes

the Lebesgue measure on $\Delta_N = \{(x_1, \dots, x_N) | x_1 < \dots < x_N\}$ and $\mu_{\mathcal{U}_N}$ is the Haar measure on the set \mathcal{U}_N of unitary matrices of size N . For any $n_1, \dots, n_p \geq 0$, one has

$$\begin{aligned} & \mathbb{E} \left[\frac{1}{N} \text{Tr}[A_1^{n_1} \circ \dots \circ A_p^{n_p}] \right] \\ &= \int_{\Delta_N^p \times \mathcal{U}_N^p} \frac{1}{N^p} \sum_{i_1, \dots, i_p=1}^N \lambda_{1,i_1}^{n_1} \dots \lambda_{p,i_p}^{n_p} f_N((\lambda_{j,i}, \Lambda_j, U_j)_{j=1, \dots, p}) \\ & \quad \times g_N((\Lambda_j, U_j)_{j=1, \dots, p}) d\mu_{\Delta_N}(\Lambda_1) \dots d\mu_{\Delta_N}(\Lambda_p) d\mu_{\mathcal{U}_N}(U_1) \dots d\mu_{\mathcal{U}_N}(U_p) \\ &= \frac{1}{N^p} \sum_{i_1, \dots, i_p=1}^N \int_{\Delta_N^p} \lambda_{1,i_1}^{n_1} \dots \lambda_{p,i_p}^{n_p} h_N((\lambda_{j,i}, \Lambda_j)_{j=1, \dots, p}) d\mu_{\Delta_N}(\Lambda_1) \dots d\mu_{\Delta_N}(\Lambda_p), \end{aligned}$$

where f_N is as in the previous step and

$$\begin{aligned} & h_N((\lambda_{j,i}, \Lambda_j)_{j=1, \dots, p}) \\ &= \int_{\mathcal{U}_N^p} f_N((\lambda_{j,i}, \Lambda_j, U_j)_{j=1, \dots, p}) g_N((\Lambda_j, U_j)_{j=1, \dots, p}) d\mu_{\mathcal{U}_N}(U_1), \dots, d\mu_{\mathcal{U}_N}(U_p). \end{aligned}$$

For any $i_1, \dots, i_p = 1, \dots, N$, we have

$$\begin{aligned} & \int_{\Delta_N^p} \lambda_{1,i_1}^{n_1} \dots \lambda_{p,i_p}^{n_p} h_N((\lambda_{j,i}, \Lambda_j)_{j=1, \dots, p}) d\mu_{\Delta_N}(\Lambda_1), \dots, d\mu_{\Delta_N}(\Lambda_p) \\ &= \int_{\mathbb{R}^p} \lambda_1^{n_1} \dots \lambda_p^{n_p} h_N^{(i_1, \dots, i_p)}(\lambda_1, \dots, \lambda_p) d\lambda_1, \dots, d\lambda_p, \end{aligned}$$

where $h_N^{(i_1, \dots, i_p)}(\lambda_1, \dots, \lambda_p)$ is obtained by integrating with respect to the variables $\lambda_{k_1}, \dots, \lambda_{k_p}$ for $k_1 \neq i_1, \dots, k_p \neq i_p$. We finally obtain

$$\mathbb{E} \left[\frac{1}{N} \text{Tr}[A_1^{n_1} \circ \dots \circ A_p^{n_p}] \right] = \int_{\mathbb{R}^p} \lambda_1^{n_1}, \dots, \lambda_p^{n_p} \bar{h}_N(\lambda_1, \dots, \lambda_p) d\lambda_1, \dots, d\lambda_p,$$

where $\bar{h}_N = \frac{1}{N^p} \sum_{i_1, \dots, i_p} h_N^{(i_1, \dots, i_p)}$. Hence, a family of random variables (a_1, \dots, a_p) as in the proposition exists and its joint distribution has density \bar{h}_N with respect to the Lebesgue measure on \mathbb{R}^p .

Step 3. We now consider the general case. Let (X_1, \dots, X_p) be a family of independent random matrices, independent of (A_1, \dots, A_p) , distributed according to the standard Gaussian measure on \mathcal{H}_N with respect to the inner product $\langle A, B \rangle = N \text{Tr}[AB]$. By the regularizing process of convolution on Hermitian space, for any $\varepsilon > 0$, the joint distribution of $(A_1^\varepsilon, \dots, A_p^\varepsilon) = (A_1 + \varepsilon X_1, \dots, A_p + \varepsilon X_p)$ has a density with respect to the Lebesgue measure. By the previous step, there exists a family of random variables $(a_1^\varepsilon, \dots, a_p^\varepsilon)$ such that $\mathbb{E}[(a_1^\varepsilon)^{n_1} \dots (a_p^\varepsilon)^{n_p}] = \mathbb{E}[\frac{1}{N} \text{Tr}[(A_1^\varepsilon)^{n_1} \circ \dots \circ (A_p^\varepsilon)^{n_p}]]$ for any $n_1, \dots, n_p \geq 0$. As ε

goes to zero, $(a_1^\varepsilon, \dots, a_p^\varepsilon)$ converges in moments to a family of random variables as in the proposition. \square

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